

```
from sklearn.svm import SVC
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split
from sklearn.metrics import
accuracy_score, classification_report, confusion_matrix, roc_auc_score, ro
c_curve, precision_score, recall_score
```

```
data=pd.read_csv("agaricus-lepiota.data",names=['poison','cap-  
shape','cap-surface','cap-color','bruises','odor','gill-  
attachment','gill-spacing','gill-size','gill-color','stalk-  
shape','stalk-root','stalk-surface-above-ring','stalk-surface-below-  
ring','stalk-color-above-ring','stalk-color-below-ring','veil-  
type','veil-color','ring-number','ring-type','spore-print-  
color','population','habitat'])
```

data

	poison	cap-shape	cap-surface	cap-color	bruises	odor	gill-
attachment \							
0	p	x	s	n	t	p	
f							
1	e	x	s	y	t	a	
f							
2	e	b	s	w	t	l	
f							
3	p	x	y	w	t	p	
f							
4	e	x	s	g	f	n	
f							
...
..							
8119	e	k	s	n	f	n	
a							
8120	e	x	s	n	f	n	
a							
8121	e	f	s	n	f	n	
a							
8122	p	k	y	n	f	y	
f							
8123	e	x	s	n	f	n	
a							
	gill-spacing	gill-size	gill-color	...	stalk-surface	below-	
ring \							
0	c	n	k	...			s

1	c	b	k	...	s
2	c	b	n	...	s
3	c	n	n	...	s
4	w	b	k	...	s
...
8119	c	b	y	...	s
8120	c	b	y	...	s
8121	c	b	n	...	s
8122	c	n	b	...	k
8123	c	b	y	...	s
stalk-color-above-ring stalk-color-below-ring veil-type veil-color \					
0		w	w	p	
w					
1		w	w	p	
w					
2		w	w	p	
w					
3		w	w	p	
w					
4		w	w	p	
w					
...
.					
8119		o	o	p	
o					
8120		o	o	p	
n					
8121		o	o	p	
o					
8122		w	w	p	
w					
8123		o	o	p	
o					
ring-number ring-type spore-print-color population habitat					
0	o	p	k	s	u
1	o	p	n	n	g
2	o	p	n	n	m

```

3          o          p          k          s          u
4          o          e          n          a          g
...      ...      ...      ...      ...      ...
8119       o          p          b          c          l
8120       o          p          b          v          l
8121       o          p          b          c          l
8122       o          e          w          v          l
8123       o          p          o          c          l

```

```
[8124 rows x 23 columns]
```

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 8124 entries, 0 to 8123
```

```
Data columns (total 23 columns):
```

#	Column	Non-Null Count	Dtype
0	poison	8124 non-null	object
1	cap-shape	8124 non-null	object
2	cap-surface	8124 non-null	object
3	cap-color	8124 non-null	object
4	bruises	8124 non-null	object
5	odor	8124 non-null	object
6	gill-attachment	8124 non-null	object
7	gill-spacing	8124 non-null	object
8	gill-size	8124 non-null	object
9	gill-color	8124 non-null	object
10	stalk-shape	8124 non-null	object
11	stalk-root	8124 non-null	object
12	stalk-surface-above-ring	8124 non-null	object
13	stalk-surface-below-ring	8124 non-null	object
14	stalk-color-above-ring	8124 non-null	object
15	stalk-color-below-ring	8124 non-null	object
16	veil-type	8124 non-null	object
17	veil-color	8124 non-null	object
18	ring-number	8124 non-null	object
19	ring-type	8124 non-null	object
20	spore-print-color	8124 non-null	object
21	population	8124 non-null	object
22	habitat	8124 non-null	object

```
dtypes: object(23)
```

```
memory usage: 1.4+ MB
```

```
data.isnull().sum()
```

```

poison          0
cap-shape       0
cap-surface     0
cap-color       0

```

```
bruises          0
odor             0
gill-attachment 0
gill-spacing     0
gill-size        0
gill-color       0
stalk-shape      0
stalk-root       0
stalk-surface-above-ring 0
stalk-surface-below-ring 0
stalk-color-above-ring 0
stalk-color-below-ring 0
veil-type        0
veil-color       0
  ring-number     0
ring-type        0
spore-print-color 0
population       0
habitat          0
dtype: int64
```

```
for column in data.columns:
    xx=data[column].value_counts()
    print(xx.index)
```

```
Index(['e', 'p'], dtype='object')
Index(['x', 'f', 'k', 'b', 's', 'c'], dtype='object')
Index(['y', 's', 'f', 'g'], dtype='object')
Index(['n', 'g', 'e', 'y', 'w', 'b', 'p', 'c', 'u', 'r'],
dtype='object')
Index(['f', 't'], dtype='object')
Index(['n', 'f', 'y', 's', 'a', 'l', 'p', 'c', 'm'], dtype='object')
Index(['f', 'a'], dtype='object')
Index(['c', 'w'], dtype='object')
Index(['b', 'n'], dtype='object')
Index(['b', 'p', 'w', 'n', 'g', 'h', 'u', 'k', 'e', 'y', 'o', 'r'],
dtype='object')
Index(['t', 'e'], dtype='object')
Index(['b', '?', 'e', 'c', 'r'], dtype='object')
Index(['s', 'k', 'f', 'y'], dtype='object')
Index(['s', 'k', 'f', 'y'], dtype='object')
Index(['w', 'p', 'g', 'n', 'b', 'o', 'e', 'c', 'y'], dtype='object')
Index(['w', 'p', 'g', 'n', 'b', 'o', 'e', 'c', 'y'], dtype='object')
Index(['p'], dtype='object')
Index(['w', 'n', 'o', 'y'], dtype='object')
Index(['o', 't', 'n'], dtype='object')
Index(['p', 'e', 'l', 'f', 'n'], dtype='object')
Index(['w', 'n', 'k', 'h', 'r', 'u', 'o', 'y', 'b'], dtype='object')
Index(['v', 'y', 's', 'n', 'a', 'c'], dtype='object')
Index(['d', 'g', 'p', 'l', 'u', 'm', 'w'], dtype='object')
```

```
data.drop(["cap-color","veil-type","stalk-root","stalk-  
shape"],axis=1,inplace=True)
```

data

	poison	cap-shape	cap-surface	bruises	odor	gill-attachment	gill-spacing \
0	p	x	s	t	p		f
1	e	x	s	t	a		f
2	e	b	s	t	l		f
3	p	x	y	t	p		f
4	e	x	s	f	n		f
...
...							
8119	e	k	s	f	n		a
8120	e	x	s	f	n		a
8121	e	f	s	f	n		a
8122	p	k	y	f	y		f
8123	e	x	s	f	n		a

	gill-size	gill-color	stalk-surface-above-ring	stalk-surface-below-ring \
0	n	k		s
1	b	k		s
2	b	n		s
3	n	n		s
4	b	k		s
...
...				
8119	b	y		s
8120	b	y		s
8121	b	n		s

```

8122      n      b      s
k
8123      b      y      s
s

      stalk-color-above-ring stalk-color-below-ring veil-color ring-
number \
0      w      w      w
0
1      w      w      w
0
2      w      w      w
0
3      w      w      w
0
4      w      w      w
0
...      ...      ...      ...
...
8119      o      o      o
0
8120      o      o      n
0
8121      o      o      o
0
8122      w      w      w
0
8123      o      o      o
0

      ring-type spore-print-color population habitat
0      p      k      s      u
1      p      n      n      g
2      p      n      n      m
3      p      k      s      u
4      e      n      a      g
...      ...      ...      ...
8119      p      b      c      l
8120      p      b      v      l
8121      p      b      c      l
8122      e      w      v      l
8123      p      o      c      l

```

[8124 rows x 19 columns]

```
data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8124 entries, 0 to 8123
Data columns (total 19 columns):

```

#	Column	Non-Null Count		Dtype
0	poison	8124	non-null	object
1	cap-shape	8124	non-null	object
2	cap-surface	8124	non-null	object
3	bruises	8124	non-null	object
4	odor	8124	non-null	object
5	gill-attachment	8124	non-null	object
6	gill-spacing	8124	non-null	object
7	gill-size	8124	non-null	object
8	gill-color	8124	non-null	object
9	stalk-surface-above-ring	8124	non-null	object
10	stalk-surface-below-ring	8124	non-null	object
11	stalk-color-above-ring	8124	non-null	object
12	stalk-color-below-ring	8124	non-null	object
13	veil-color	8124	non-null	object
14	ring-number	8124	non-null	object
15	ring-type	8124	non-null	object
16	spore-print-color	8124	non-null	object
17	population	8124	non-null	object
18	habitat	8124	non-null	object

dtypes: object(19)
memory usage: 1.2+ MB

data.describe()

	poison	cap-shape	cap-surface	bruises	odor	gill-attachment	\
count	8124	8124	8124	8124	8124	8124	
unique	2	6	4	2	9	2	
top	e	x	y	f	n	f	
freq	4208	3656	3244	4748	3528	7914	

	gill-spacing	gill-size	gill-color	stalk-surface-above-ring	\
count	8124	8124	8124	8124	
unique	2	2	12	4	
top	c	b	b	s	
freq	6812	5612	1728	5176	

	stalk-surface-below-ring	stalk-color-above-ring	stalk-color-below-ring	\
count	8124	8124	8124	
unique	4	9	9	
top	s	w	w	
freq	4936	4464	4384	

	veil-color	ring-number	ring-type	spore-print-color	population
--	------------	-------------	-----------	-------------------	------------

habitat					
count	8124	8124	8124	8124	8124
unique	4	3	5	9	6
top	w	o	p	w	v
d					
freq	7924	7488	3968	2388	4040
3148					

data.dtypes

poison	object
cap-shape	object
cap-surface	object
bruises	object
odor	object
gill-attachment	object
gill-spacing	object
gill-size	object
gill-color	object
stalk-surface-above-ring	object
stalk-surface-below-ring	object
stalk-color-above-ring	object
stalk-color-below-ring	object
veil-color	object
ring-number	object
ring-type	object
spore-print-color	object
population	object
habitat	object
dtype:	object

```
le=LabelEncoder()
for column in data.columns:
    data[column]=le.fit_transform(data[column])
```

data.head()

	poison	cap-shape	cap-surface	bruises	odor	gill-attachment	\
0	1	5	2	1	6	1	
1	0	5	2	1	0	1	
2	0	0	2	1	3	1	
3	1	5	3	1	6	1	
4	0	5	2	0	5	1	

	gill-spacing	gill-size	gill-color	stalk-surface-above-ring	\
0	0	1	4	2	
1	0	0	4	2	
2	0	0	5	2	

3	0	1	5	2
4	1	0	4	2

	stalk-surface-below-ring	stalk-color-above-ring	stalk-color-below-ring \
--	--------------------------	------------------------	--------------------------

0	2	7
7		
1	2	7
7		
2	2	7
7		
3	2	7
7		
4	2	7
7		

	veil-color	ring-number	ring-type	spore-print-color	population
--	------------	-------------	-----------	-------------------	------------

0	2	1	4	2	3
5					
1	2	1	4	3	2
1					
2	2	1	4	3	2
3					
3	2	1	4	2	3
5					
4	2	1	0	3	0
1					

```
x=data.iloc[:,1:].values
y=data.iloc[:,0].values
```

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
```

```
cl1=SVC(kernel="rbf",random_state=0,probability=True)
cl1.fit(x_train,y_train)
```

```
SVC(probability=True, random_state=0)
```

```
cl1.score(x_train,y_train)
```

```
0.9832281889521465
```

```
ypred1=cl1.predict(x_test)
```

```
acc1=accuracy_score(y_test,ypred1)
```

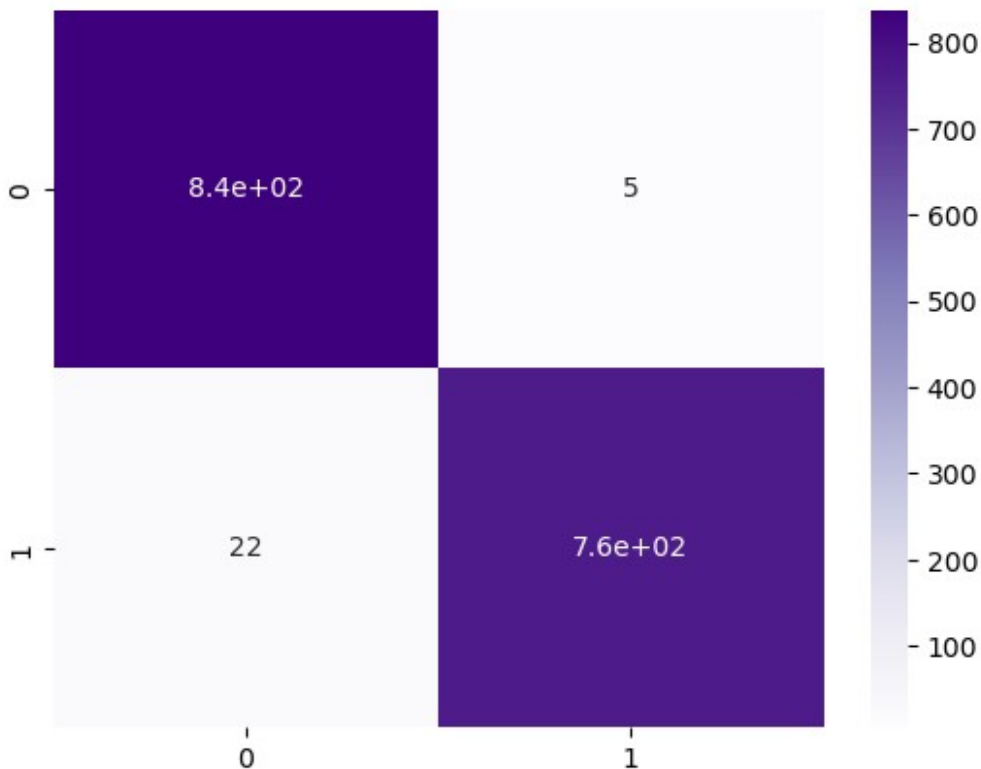
```
acc1
```

```
0.9833846153846154
```

```

print("pscore",precision_score(y_test,ypred1))
pscore 0.9934640522875817
print("rscore",recall_score(y_test,ypred1))
rscore 0.9718670076726342
cm1=confusion_matrix(y_test,ypred1)
cm1
array([[838,    5],
       [ 22, 760]], dtype=int64)
sns.heatmap(cm1,cmap='Purples',annot=True)
<Axes: >

```

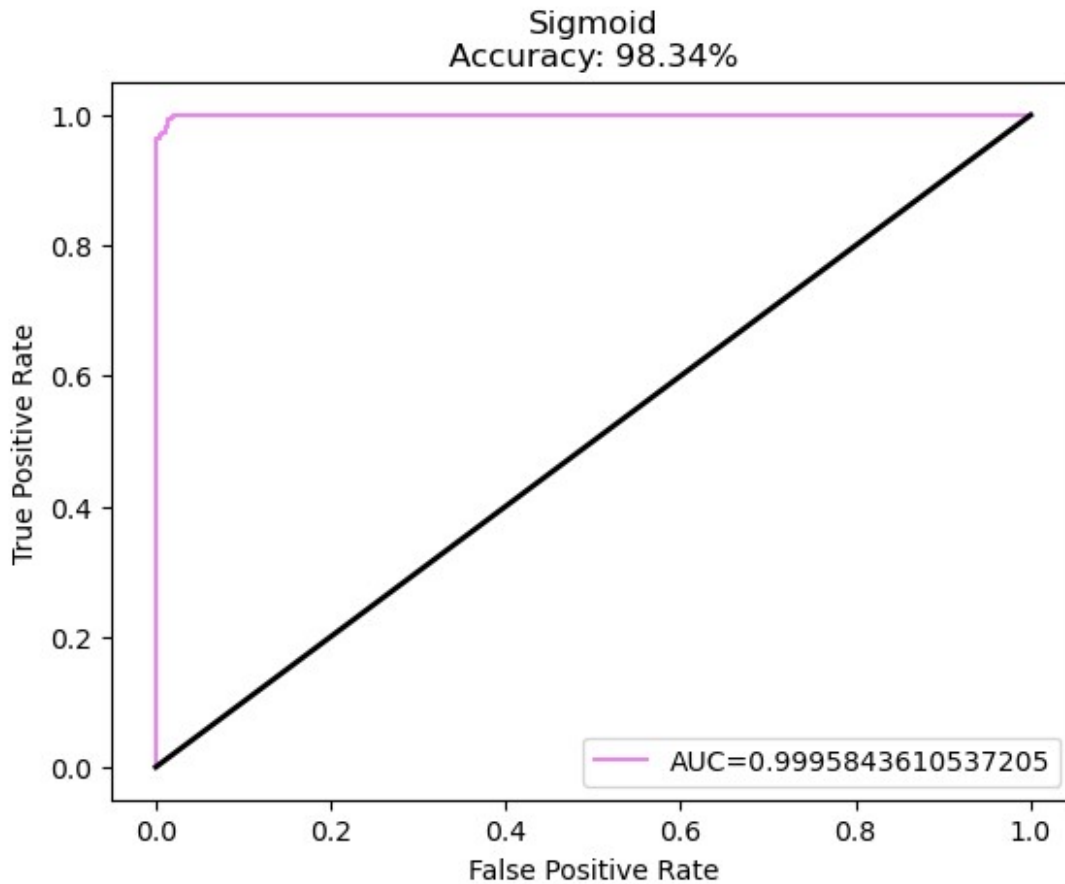


```

import matplotlib.pyplot as plt
from sklearn import metrics
y_pred_proba = c1.predict_proba(x_test)[:,1]
fpr, tpr, _ = metrics.roc_curve(y_test, y_pred_proba)
auc = metrics.roc_auc_score(y_test, y_pred_proba)
plt.plot(fpr,tpr,label="AUC="+str(auc),color='violet')
plt.legend(loc=4)

```

```
plt.plot([0, 1], [0, 1], color='black', lw=2)
plt.ylabel('True Positive Rate')
plt.xlabel('False Positive Rate')
plt.title('Sigmoid\nAccuracy: {:.2f}%'.format(acc1 * 100))
plt.show()
```



```
cl2=SVC(kernel="linear",random_state=0,probability=True)
cl2.fit(x_train,y_train)

SVC(kernel='linear', probability=True, random_state=0)
cl2.score(x_train,y_train)
0.9524542237267272

ypred2=cl2.predict(x_test)

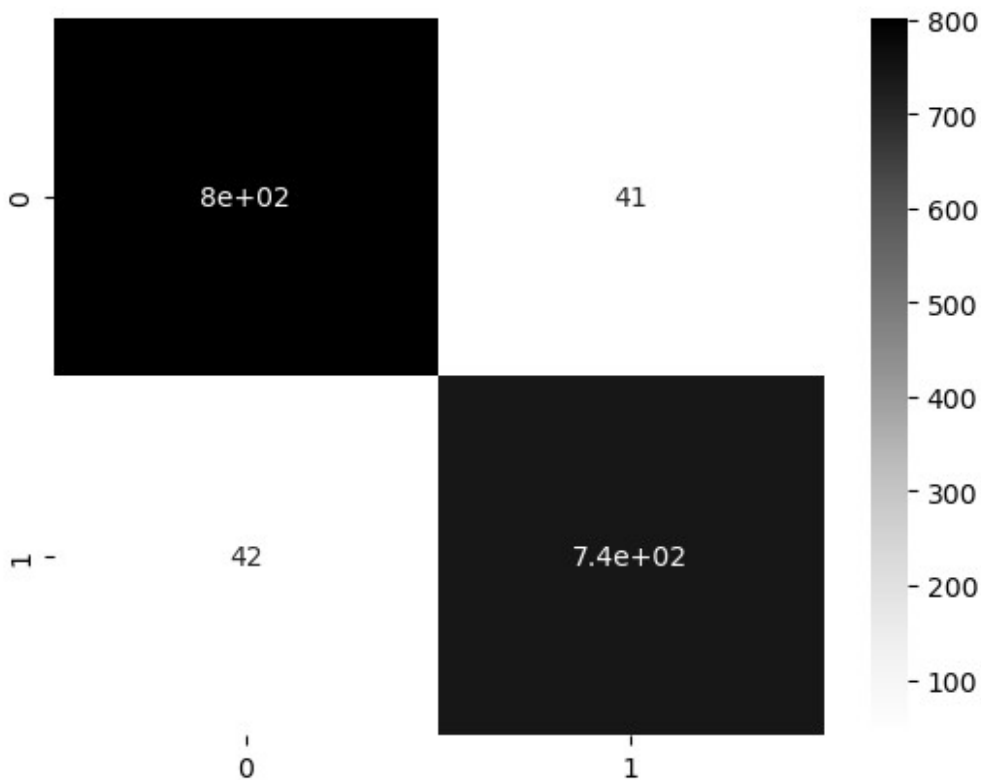
acc2=accuracy_score(y_test,ypred2)
acc2
0.9489230769230769

print("pscore",precision_score(y_test,ypred2))
```

```

pscore 0.9475032010243278
print("pscore",precision_score(y_test,ypred2))
pscore 0.9475032010243278
cm2=confusion_matrix(y_test,ypred2)
cm2
array([[802,  41],
       [ 42, 740]], dtype=int64)
sns.heatmap(cm2,cmap='Greys',annot=True)
<Axes: >

```

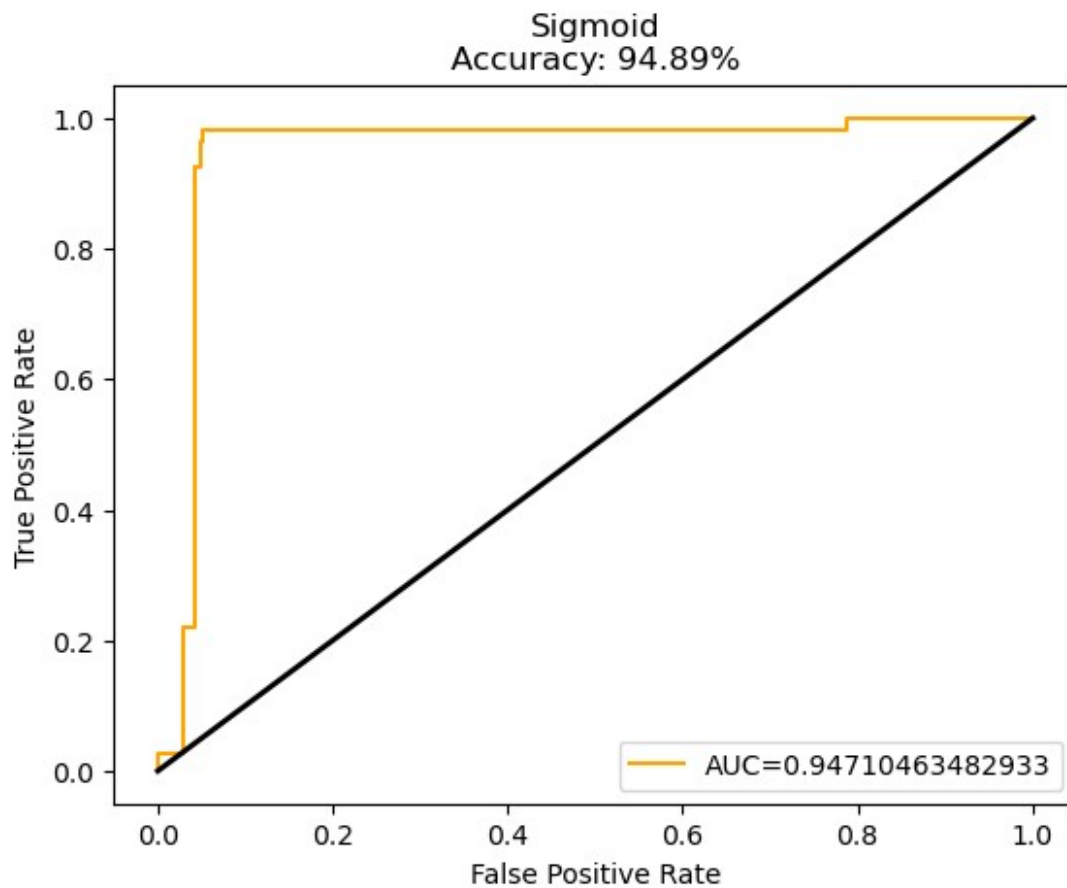


```

import matplotlib.pyplot as plt
from sklearn import metrics
y_pred_proba = cl2.predict_proba(x_test)[:,:1]
fpr, tpr, _ = metrics.roc_curve(y_test, y_pred_proba)
auc = metrics.roc_auc_score(y_test, y_pred_proba)
plt.plot(fpr,tpr,label="AUC="+str(auc),color='orange')
plt.legend(loc=4)
plt.plot([0, 1], [0, 1], color='black', lw=2)
plt.ylabel('True Positive Rate')
plt.xlabel('False Positive Rate')

```

```
plt.title('Sigmoid\nAccuracy: {:.2f}%'.format(acc2* 100))
plt.show()
```



```
cl3=SVC(kernel="poly",random_state=0,probability=True)
cl3.fit(x_train,y_train)

SVC(kernel='poly', probability=True, random_state=0)

cl3.score(x_train,y_train)
0.9956916448684413

ypred3=cl3.predict(x_test)

acc3=accuracy_score(y_test,ypred3)
acc3
0.9969230769230769

print("pscore",precision_score(y_test,ypred3))
pscore 0.9949044585987261
```

```
print("rscore",precision_score(y_test,ypred3))
```

```
rscore 0.9949044585987261
```

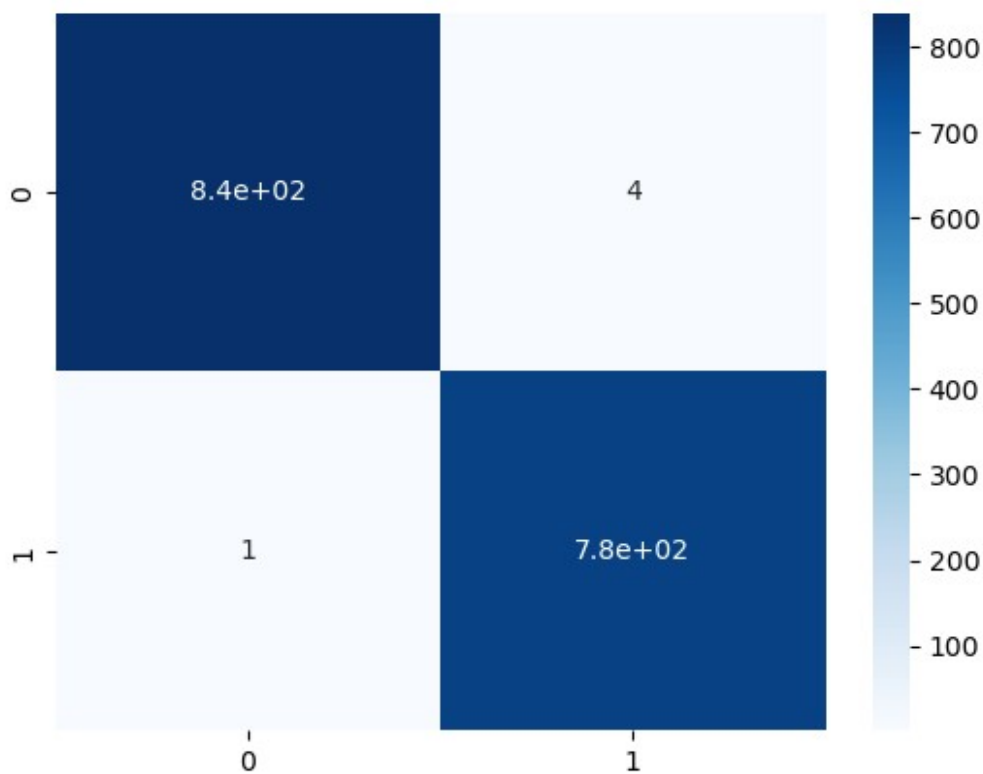
```
cm3=confusion_matrix(y_test,ypred3)
```

```
cm3
```

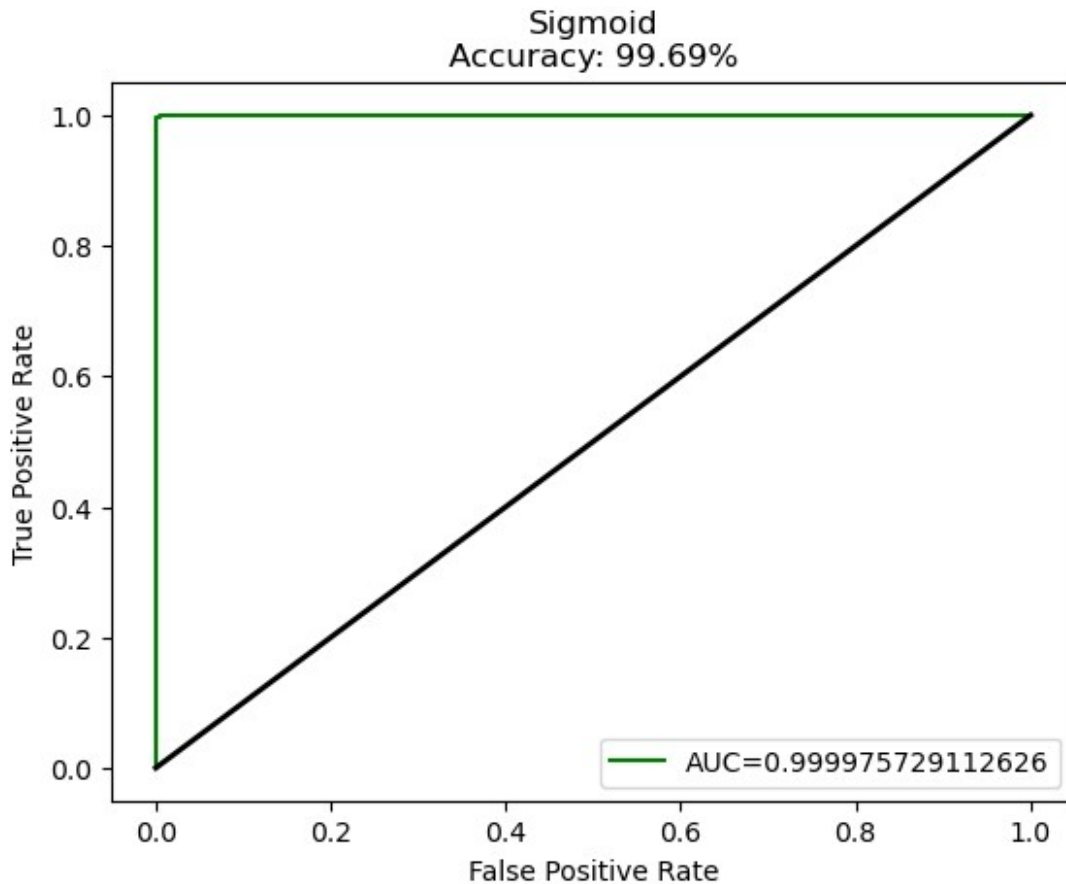
```
array([[839,  4],  
       [ 1, 781]], dtype=int64)
```

```
sns.heatmap(cm3,cmap='Blues',annot=True)
```

```
<Axes: >
```



```
import matplotlib.pyplot as plt
from sklearn import metrics
y_pred_proba = cl3.predict_proba(x_test)[::,1]
fpr, tpr, _ = metrics.roc_curve(y_test, y_pred_proba)
auc = metrics.roc_auc_score(y_test, y_pred_proba)
plt.plot(fpr,tpr,label="AUC="+str(auc),color='green')
plt.legend(loc=4)
plt.plot([0, 1], [0, 1], color='black', lw=2)
plt.ylabel('True Positive Rate')
plt.xlabel('False Positive Rate')
plt.title('Sigmoid\nAccuracy: {:.2f}%'.format(acc3 * 100))
plt.show()
```



```
cl4=SVC(kernel="sigmoid",random_state=0,probability=True)
cl4.fit(x_train,y_train)

SVC(kernel='sigmoid', probability=True, random_state=0)
cl4.score(x_train,y_train)
0.4756116325588552

ypred4=cl4.predict(x_test)
acc4=accuracy_score(y_test,ypred4)
acc4
0.47015384615384614

print("pscore",precision_score(y_test,ypred4))
pscore 0.4496815286624204

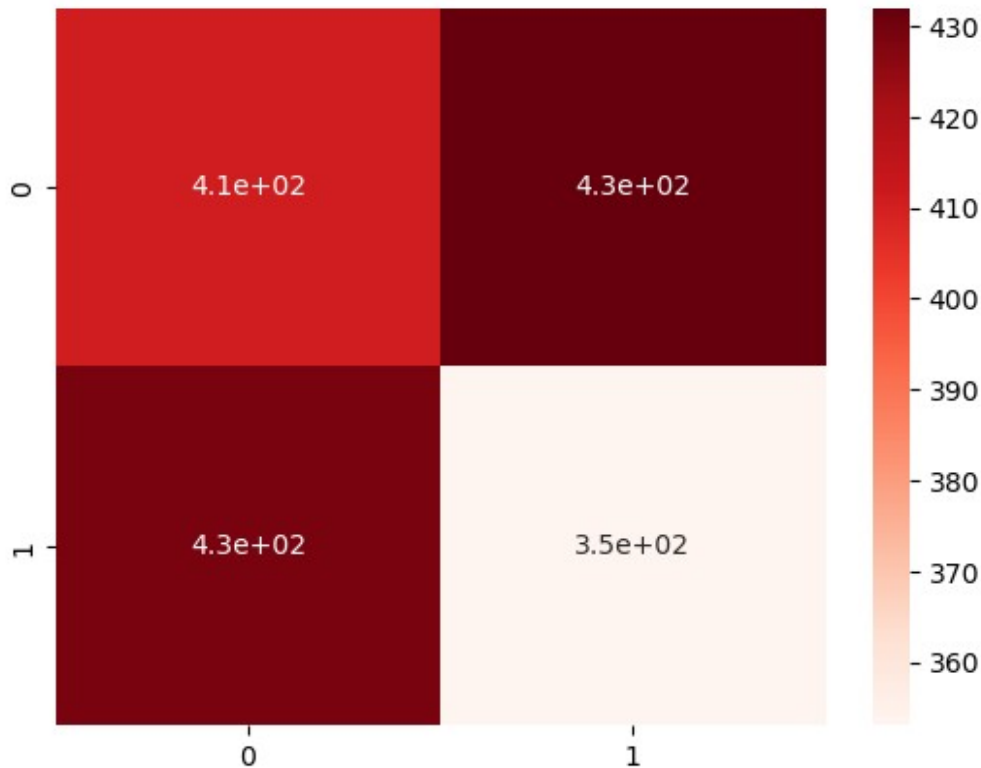
print("rscore",precision_score(y_test,ypred4))
rscore 0.4496815286624204
```

```

cm4=confusion_matrix(y_test,ypred4)
cm4
array([[411, 432],
       [429, 353]], dtype=int64)

sns.heatmap(cm4,cmap='Reds',annot=True)
<Axes: >

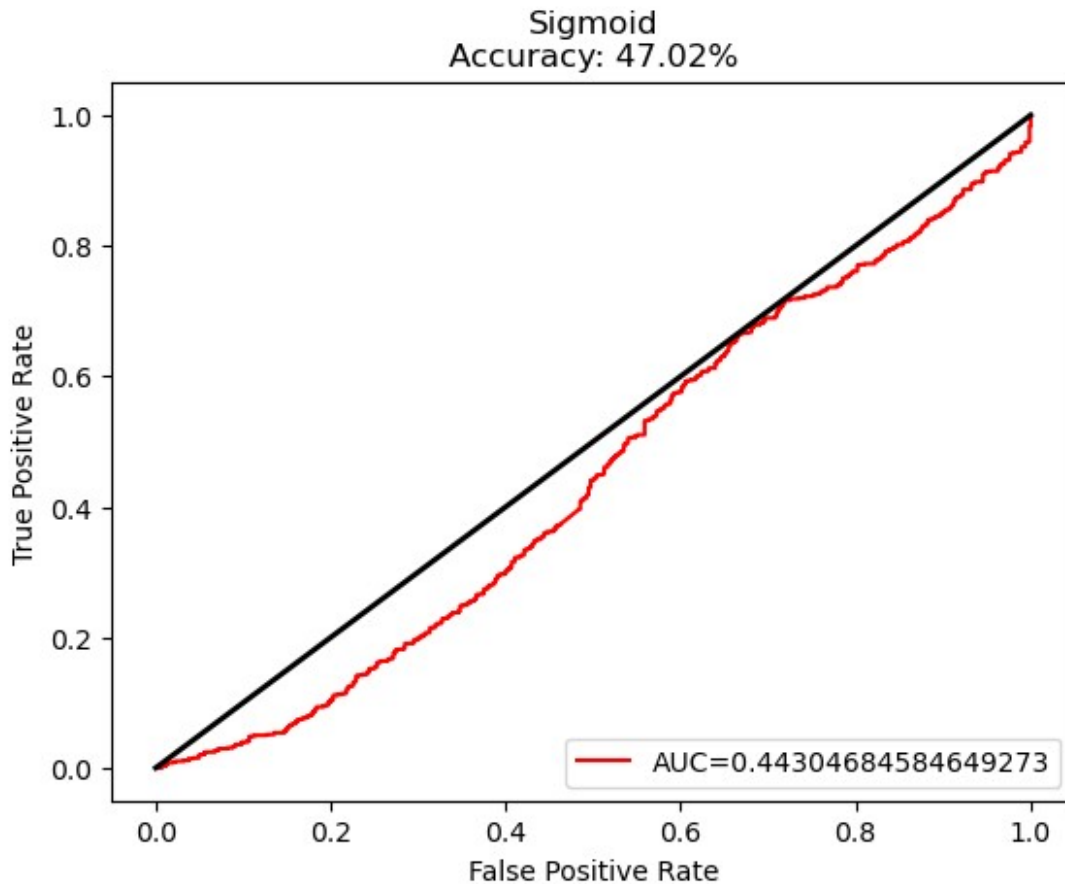
```



```

import matplotlib.pyplot as plt
from sklearn import metrics
y_pred_proba = cl4.predict_proba(x_test)[::,1]
fpr, tpr, _ = metrics.roc_curve(y_test, y_pred_proba)
auc = metrics.roc_auc_score(y_test, y_pred_proba)
plt.plot(fpr,tpr,label="AUC="+str(auc),color='red')
plt.legend(loc=4)
plt.plot([0, 1], [0, 1], color='black', lw=2)
plt.ylabel('True Positive Rate')
plt.xlabel('False Positive Rate')
plt.title('Sigmoid\nAccuracy: {:.2f}%'.format(acc4 * 100))
plt.show()

```

```
pip install nbconvert
```

Requirement already satisfied: nbconvert in c:\users\new\anaconda3\lib\site-packages (6.5.4)

Requirement already satisfied: lxml in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (4.9.3)

Requirement already satisfied: beautifulsoup4 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (4.12.2)

Requirement already satisfied: bleach in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (4.1.0)

Requirement already satisfied: defusedxml in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (0.7.1)

Requirement already satisfied: entrypoints>=0.2.2 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (0.4)

Requirement already satisfied: jinja2>=3.0 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (3.1.2)

Requirement already satisfied: jupyter-core>=4.7 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (5.3.0)

Requirement already satisfied: jupyterlab-pygments in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (0.1.2)

Requirement already satisfied: MarkupSafe>=2.0 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (2.1.1)

Requirement already satisfied: mistune<2,>=0.8.1 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (0.8.4)

Requirement already satisfied: nbclient>=0.5.0 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (0.5.13)

Requirement already satisfied: nbformat>=5.1 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (5.9.2)

Requirement already satisfied: packaging in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (23.1)

Requirement already satisfied: pandocfilters>=1.4.1 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (1.5.0)

Requirement already satisfied: pygments>=2.4.1 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (2.15.1)

Requirement already satisfied: tinycss2 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (1.2.1)

Requirement already satisfied: traitlets>=5.0 in c:\users\new\anaconda3\lib\site-packages (from nbconvert) (5.7.1)

Requirement already satisfied: platformdirs>=2.5 in c:\users\new\anaconda3\lib\site-packages (from jupyter-core>=4.7->nbconvert) (3.10.0)

Requirement already satisfied: pywin32>=300 in c:\users\new\anaconda3\lib\site-packages (from jupyter-core>=4.7->nbconvert) (305.1)

Requirement already satisfied: jupyter-client>=6.1.5 in c:\users\new\anaconda3\lib\site-packages (from nbclient>=0.5.0->nbconvert) (7.4.9)

Requirement already satisfied: nest-asyncio in c:\users\new\anaconda3\lib\site-packages (from nbclient>=0.5.0->nbconvert) (1.5.6)

Requirement already satisfied: fastjsonschema in c:\users\new\anaconda3\lib\site-packages (from nbformat>=5.1->nbconvert) (2.16.2)

Requirement already satisfied: jsonschema>=2.6 in c:\users\new\anaconda3\lib\site-packages (from nbformat>=5.1->nbconvert) (4.17.3)

Requirement already satisfied: soupsieve>1.2 in c:\users\new\anaconda3\lib\site-packages (from beautifulsoup4->nbconvert) (2.4)

Requirement already satisfied: six>=1.9.0 in c:\users\new\anaconda3\lib\site-packages (from bleach->nbconvert) (1.16.0)

Requirement already satisfied: webencodings in c:\users\new\anaconda3\lib\site-packages (from bleach->nbconvert) (0.5.1)

Requirement already satisfied: attrs>=17.4.0 in c:\users\new\anaconda3\lib\site-packages (from jsonschema>=2.6->nbformat>=5.1->nbconvert) (23.2.0)

Requirement already satisfied: pyrsistent!=0.17.0,!0.17.1,!0.17.2,>=0.14.0 in c:\users\new\anaconda3\lib\site-packages (from jsonschema>=2.6->nbformat>=5.1->nbconvert) (0.18.0)

Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\new\anaconda3\lib\site-packages (from jupyter-client>=6.1.5->nbclient>=0.5.0->nbconvert) (2.8.2)

Requirement already satisfied: pyzmq>=23.0 in c:\users\new\anaconda3\lib\site-packages (from jupyter-client>=6.1.5->nbclient>=0.5.0->nbconvert) (23.2.0)

Requirement already satisfied: tornado>=6.2 in c:\users\new\anaconda3\lib\site-packages (from jupyter-client>=6.1.5->nbclient>=0.5.0->nbconvert) (6.2.0)

```
>nbconvert) (6.3.2)
```

Note: you may need to restart the kernel to use updated packages.