

```
# Importing Libraries:
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
dataset=pd.read_csv(r"/content/kidney_disease.csv")
```

```
# Top 5 records:
dataset.head()
```

	id	age	bp	sg	al	su	rbc	pc	pcc	ba	bgr	bu	sc	sod	pot	hemo	pcv	wc	rc	htn
0	0	48.0	80.0	1.020	1.0	0.0	NaN	normal	notpresent	notpresent	121.0	36.0	1.2	NaN	NaN	15.4	44	7800	5.2	yes
1	1	7.0	50.0	1.020	4.0	0.0	NaN	normal	notpresent	notpresent	NaN	18.0	0.8	NaN	NaN	11.3	38	6000	NaN	no
2	2	62.0	80.0	1.010	2.0	3.0	normal	normal	notpresent	notpresent	423.0	53.0	1.8	NaN	NaN	9.6	31	7500	NaN	no
3	3	48.0	70.0	1.005	4.0	0.0	normal	abnormal	present	notpresent	117.0	56.0	3.8	111.0	2.5	11.2	32	6700	3.9	yes
4	4	51.0	80.0	1.010	2.0	0.0	normal	normal	notpresent	notpresent	106.0	26.0	1.4	NaN	NaN	11.6	35	7300	4.6	no

Saved successfully!



```
# Dropping unnecessary feature :
dataset = dataset.drop('id', axis=1)
```

```
# Shape of dataset:
dataset.shape
```

```
(400, 25)
```

```
# Cheaking Missing (NaN) Values:  
dataset.isnull().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

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```
# Description:  
dataset.describe()
```

	age	bp	sg	al	su	bgr	bu	sc	sod	pot	
count	391.000000	388.000000	353.000000	354.000000	351.000000	356.000000	381.000000	383.000000	313.000000	312.000000	348.00
mean	51.483376	76.469072	1.017408	1.016949	0.450142	148.036517	57.425722	3.072454	137.528754	4.627244	12.52
std	17.169714	13.683637	0.005717	1.352679	1.099191	79.281714	50.503006	5.741126	10.408752	3.193904	2.91
min	2.000000	50.000000	1.005000	0.000000	0.000000	22.000000	1.500000	0.400000	4.500000	2.500000	3.10
25%	42.000000	70.000000	1.010000	0.000000	0.000000	99.000000	27.000000	0.900000	135.000000	3.800000	10.30

Datatypes:
dataset.dtypes

```

age      float64
bp       float64
sg       float64
al       float64
su       float64
rbc      object
pc       object
pcc      object
ba       object
bgr      float64
bu       float64
sc       float64
sod      float64

```

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```

wc       object
rc       object
htn      object
dm       object
cad      object
appet    object
pe       object
ane      object
classification
dtype: object

```

```
dataset.head()
```

	age	bp	sg	al	su	rbc	pc	pcc	ba	bgr	...	pcv	wc	rc	htn	dm	cad	appet	pe	ane
0	48.0	80.0	1.020	1.0	0.0	NaN	normal	notpresent	notpresent	121.0	...	44	7800	5.2	yes	yes	no	good	no	no
1	7.0	50.0	1.020	4.0	0.0	NaN	normal	notpresent	notpresent	NaN	...	38	6000	NaN	no	no	no	good	no	no
2	62.0	80.0	1.010	2.0	3.0	normal	normal	notpresent	notpresent	423.0	...	31	7500	NaN	no	yes	no	poor	no	yes
3	48.0	70.0	1.005	4.0	0.0	normal	abnormal	present	notpresent	117.0	...	32	6700	3.9	yes	no	no	poor	yes	yes
4	51.0	80.0	1.010	2.0	0.0	normal	normal	notpresent	notpresent	106.0	...	35	7300	4.6	no	no	no	good	no	no

5 rows x 25 columns

```
dataset['rbc'].value_counts()
```

```
normal      201
abnormal     47
Name: rbc, dtype: int64
```

```
dataset['rbc'] = dataset['rbc'].replace(to_replace = {'normal' : 0, 'abnormal' : 1})
```

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```
normal      201
abnormal     76
Name: pc, dtype: int64
```

```
dataset['pc'] = dataset['pc'].replace(to_replace = {'normal' : 0, 'abnormal' : 1})
```

```
dataset['pcc'].value_counts()
```

```
notpresent    354
```

```
present      42
Name: pcc, dtype: int64
```

```
dataset['pcc'] = dataset['pcc'].replace(to_replace = {'notpresent':0, 'present':1})
```

```
dataset['ba'].value_counts()
```

```
notpresent    374
present        22
Name: ba, dtype: int64
```

```
dataset['ba'] = dataset['ba'].replace(to_replace = {'notpresent':0, 'present':1})
```

```
dataset['htn'].value_counts()
```

```
no      251
yes     147
Name: htn, dtype: int64
```

```
dataset['htn'] = dataset['htn'].replace(to_replace = {'yes' : 1, 'no' : 0})
```

```
dataset['dm'].value_counts()
```

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```
\tno      3
\tyes     2
yes        1
Name: dm, dtype: int64
```

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

```
dataset['dm'] = dataset['dm'].replace(to_replace = {'yes' : 1, 'no' : 0})
```

```
dataset['cad'].value_counts()
```

```
no      362
yes      34
\tno     2
Name: cad, dtype: int64
```

```
dataset['cad'] = dataset['cad'].replace(to_replace = {'\tno':'no'})
```

```
dataset['cad'] = dataset['cad'].replace(to_replace = {'yes' : 1, 'no' : 0})
```

```
dataset['appet'].unique()
```

```
array(['good', 'poor', nan], dtype=object)
```

```
dataset['appet'] = dataset['appet'].replace(to_replace={'good':1,'poor':0,'no':np.nan})
```

```
dataset['pe'].value_counts()
```

```
no      323
yes      76
Name: pe, dtype: int64
```

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```
dataset['pe'] = dataset['pe'].replace(to_replace = {'yes' : 1, 'no' : 0})
```

```
dataset['ane'].value_counts()
```

```
no      339
yes      60
Name: ane, dtype: int64
```

```
dataset['ane'] = dataset['ane'].replace(to_replace = {'yes' : 1, 'no' : 0})
```

```
dataset['classification'].value_counts()
```

```
ckd      248
notckd   150
ckd\t      2
Name: classification, dtype: int64
```

```
dataset['classification'] = dataset['classification'].replace(to_replace={'ckd\t':'ckd'})
```

```
dataset["classification"] = [1 if i == "ckd" else 0 for i in dataset["classification"]]
```

```
dataset.head()
```

	age	bp	sg	al	su	rbc	pc	pcc	ba	bgr	...	pcv	wc	rc	htn	dm	cad	appet	pe	ane	classification
0	48.0	80.0	1.020	1.0	0.0	NaN	0.0	0.0	0.0	121.0	...	44	7800	5.2	1.0	1.0	0.0	1.0	0.0	0.0	1
1	7.0	50.0	1.020	4.0	0.0	NaN	0.0	0.0	0.0	NaN	...	38	6000	NaN	0.0	0.0	0.0	1.0	0.0	0.0	1
2	62.0	80.0	1.010	2.0	3.0	0.0	0.0	0.0	0.0	423.0	...	31	7500	NaN	0.0	1.0	0.0	0.0	0.0	1.0	1
3	48.0	70.0	1.005	4.0	0.0	0.0	1.0	1.0	0.0	117.0	...	32	6700	3.9	1.0	0.0	0.0	0.0	1.0	1.0	1
4	51.0	80.0	1.010	2.0	0.0	0.0	0.0	0.0	0.0	106.0	...	35	7300	4.6	0.0	0.0	0.0	1.0	0.0	0.0	1

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```
# Datatypes:
dataset.dtypes
```

```
age      float64
bp      float64
sg      float64
al      float64
su      float64
```

```
rbc      float64
pc        float64
pcc       float64
ba        float64
bgr       float64
bu        float64
sc        float64
sod       float64
pot       float64
hemo      float64
pcv       object
wc        object
rc        object
htn       float64
dm        float64
cad       float64
appet     float64
pe        float64
ane       float64
classification  int64
dtype: object
```

```
dataset['pcv'] = pd.to_numeric(dataset['pcv'], errors='coerce')
dataset['wc'] = pd.to_numeric(dataset['wc'], errors='coerce')
dataset['rc'] = pd.to_numeric(dataset['rc'], errors='coerce')
```

Saved successfully!



```
age      float64
bp       float64
sg       float64
al       float64
su       float64
rbc      float64
pc       float64
pcc      float64
ba       float64
bgr      float64
```



```
bu          float64
sc          float64
sod         float64
pot         float64
hemo        float64
pcv         float64
wc          float64
rc          float64
htn         float64
dm          float64
cad         float64
appet       float64
pe          float64
ane         float64
classification  int64
dtype: object
```

```
# Description:
dataset.describe()
```

Saved successfully!



age	bp	sg	al	su	rbc	pc	pcc	ba	bgr	...
-----	----	----	----	----	-----	----	-----	----	-----	-----

```
# Cheaking Missing (NaN) Values:
```

```
dataset.isnull().sum().sort_values(ascending=False)
```

rbc	152
rc	131
wc	106
pot	88
sod	87
pcv	71
pc	65
hemo	52
su	49
sg	47
al	46
bgr	44
bu	19
sc	17
bp	12
age	9
ba	4
pcc	4
htn	2
dm	2
cad	2

Saved successfully!



```
classification    0
dtype: int64
```

```
dataset.columns
```

```
Index(['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'],
      dtype='object')
```

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

```
for feature in features:  
    dataset[feature] = dataset[feature].fillna(dataset[feature].median())
```

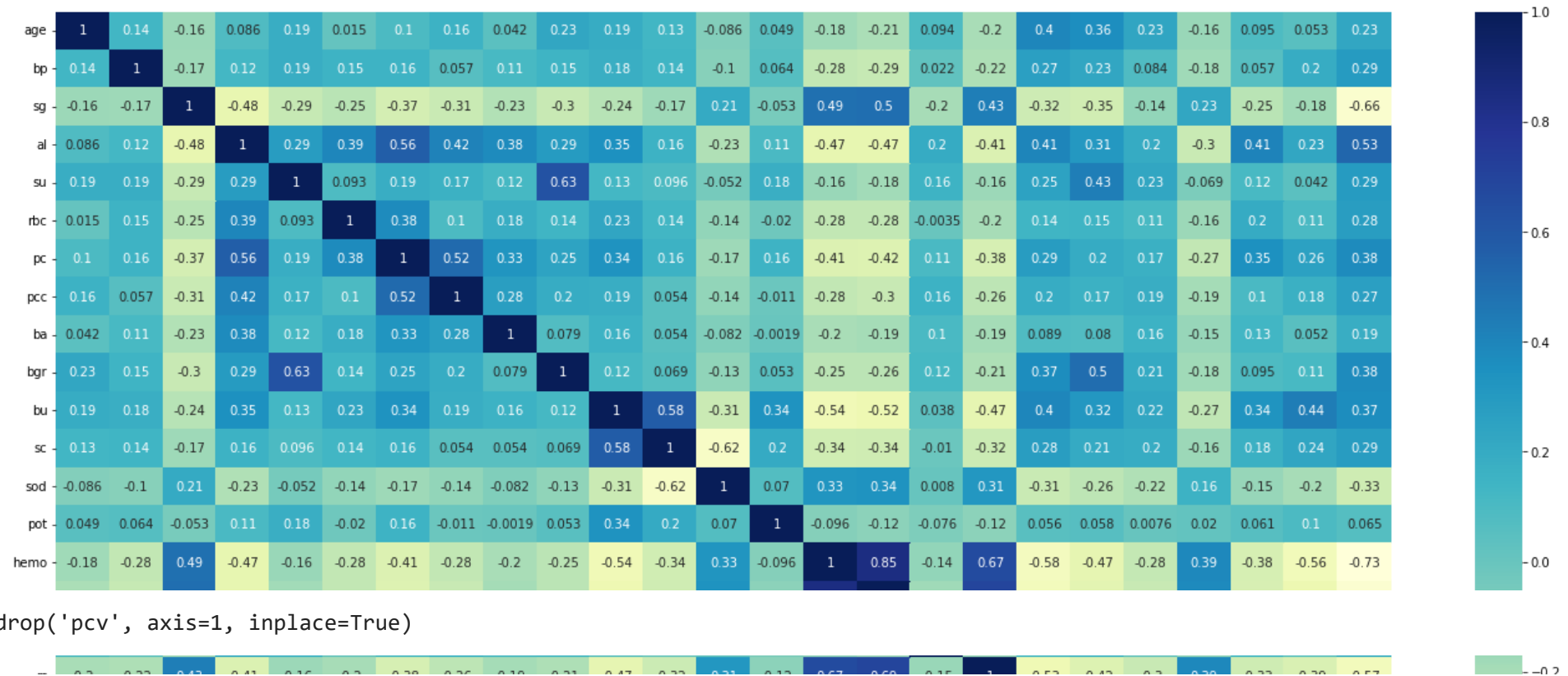
```
dataset.isnull().any().sum()
```

```
0
```

```
plt.figure(figsize=(24,14))  
sns.heatmap(dataset.corr(), annot=True, cmap='YlGnBu')  
plt.show()
```

Saved successfully!





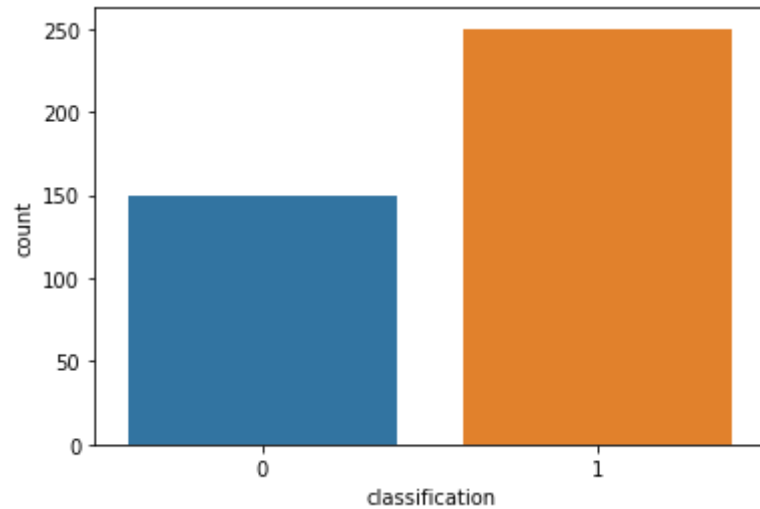
	age	bp	sg	al	su	rbc	pc	pcc	ba	bgr	...	hemo	wc	rc	htn	dm	cad	appet	pe	ane	classification
						0.0	0.0	0.0	0.0	121.0	...	15.4	7800.0	5.2	1.0	1.0	0.0	1.0	0.0	0.0	1
1	7.0	50.0	1.020	4.0	0.0	0.0	0.0	0.0	0.0	121.0	...	11.3	6000.0	4.8	0.0	0.0	0.0	1.0	0.0	0.0	1
2	62.0	80.0	1.010	2.0	3.0	0.0	0.0	0.0	0.0	423.0	...	9.6	7500.0	4.8	0.0	1.0	0.0	0.0	0.0	1.0	1
3	48.0	70.0	1.005	4.0	0.0	0.0	1.0	1.0	0.0	117.0	...	11.2	6700.0	3.9	1.0	0.0	0.0	0.0	1.0	1.0	1
4	51.0	80.0	1.010	2.0	0.0	0.0	0.0	0.0	0.0	106.0	...	11.6	7300.0	4.6	0.0	0.0	0.0	1.0	0.0	0.0	1

5 rows × 24 columns

```
# Target feature:
```

```
sns.countplot(dataset['classification'])
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg:  
FutureWarning  
<matplotlib.axes._subplots.AxesSubplot at 0x7f82460d0790>
```



```
# Independent and Dependent Feature:
```

```
X = dataset.iloc[:, :-1]
```

```
y = dataset.iloc[:, -1]
```

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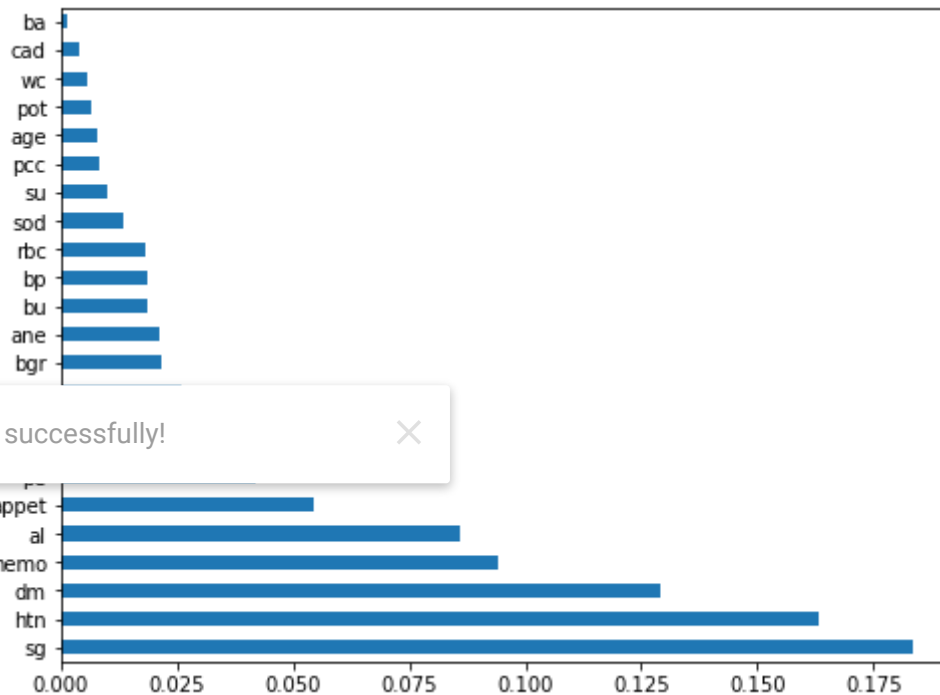


	age	bp	sg	al	su	rbc	pc	pcc	ba	bgr	...	pot	hemo	wc	rc	htn	dm	cad	appet	pe	ane
0	48.0	80.0	1.020	1.0	0.0	0.0	0.0	0.0	0.0	121.0	...	4.4	15.4	7800.0	5.2	1.0	1.0	0.0	1.0	0.0	0.0

Feature Importance:

```
from sklearn.ensemble import ExtraTreesClassifier
import matplotlib.pyplot as plt
model=ExtraTreesClassifier()
model.fit(X,y)
```

```
plt.figure(figsize=(8,6))
ranked_features=pd.Series(model.feature_importances_,index=X.columns)
ranked_features.nlargest(24).plot(kind='barh')
plt.show()
```



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```
ranked_features.nlargest(8).index
```

```
Index(['sg', 'htn', 'dm', 'hemo', 'al', 'appet', 'pe', 'pc'], dtype='object')
```

```
X = dataset[['sg', 'htn', 'hemo', 'dm', 'al', 'appet', 'rc', 'pc']]
X.head()
```

	sg	htn	hemo	dm	al	appet	rc	pc
0	1.020	1.0	15.4	1.0	1.0	1.0	5.2	0.0
1	1.020	0.0	11.3	0.0	4.0	1.0	4.8	0.0
2	1.010	0.0	9.6	1.0	2.0	0.0	4.8	0.0
3	1.005	1.0	11.2	0.0	4.0	0.0	3.9	1.0
4	1.010	0.0	11.6	0.0	2.0	1.0	4.6	0.0

```
X.tail()
```

	sg	htn	hemo	dm	al	appet	rc	pc
395	1.020	0.0	15.7	0.0	0.0	1.0	4.9	0.0
396	1.025	0.0	16.5	0.0	0.0	1.0	6.2	0.0
397	1.020	0.0	15.8	0.0	0.0	1.0	5.4	0.0
398	1.020	0.0	15.8	0.0	0.0	1.0	5.9	0.0
399	1.025	0.0	15.8	0.0	0.0	1.0	6.1	0.0

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```
y.head()
```

```
0    1
1    1
2    1
3    1
```

```

4      1
      Name: classification. dtype: int64

# Train Test Split:
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test = train_test_split(X,y, test_size=0.3, random_state=33)

print(X_train.shape)
print(X_test.shape)

(280, 8)
(120, 8)

# Importing Performance Metrics:
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report

# RandomForestClassifier:
from sklearn.ensemble import RandomForestClassifier
RandomForest = RandomForestClassifier()
RandomForest = RandomForest.fit(X_train,y_train)

# Predictions:
y_pred = RandomForest.predict(X_test)
print(y_pred)

Saved successfully!
print(accuracy_score(y_test,y_pred))
print(confusion_matrix(y_test,y_pred))
print(classification_report(y_test,y_pred))

[1 0 0 1 1 1 1 1 1 1 0 0 1 0 0 1 0 0 1 1 0 1 1 0 1 1 1 1 1 0 0 0 0 0 0 1 0
 1 1 1 1 0 1 0 0 1 0 0 1 1 0 1 0 1 1 1 1 0 1 1 0 0 0 0 1 0 1 0 0 1 1 0 0
 0 0 1 0 0 0 0 0 0 1 0 0 1 0 1 1 1 1 1 0 0 1 1 1 0 1 1 1 1 1 1 1 0 1 1 0
 1 1 1 0 1 0 0 0 1]
Accuracy: 0.975
[[55  3]
 [ 0 62]]

```


	precision	recall	f1-score	support
0	1.00	0.95	0.97	58
1	0.95	1.00	0.98	62
accuracy			0.97	120
macro avg	0.98	0.97	0.97	120
weighted avg	0.98	0.97	0.97	120

```
# AdaBoostClassifier:
from sklearn.ensemble import AdaBoostClassifier
AdaBoost = AdaBoostClassifier()
AdaBoost = AdaBoost.fit(X_train,y_train)
```

```
# Predictions:
y_pred = AdaBoost.predict(X_test)
print(y_pred)
# Performance:
print('Accuracy:', accuracy_score(y_test,y_pred))
print(confusion_matrix(y_test,y_pred))
print(classification_report(y_test,y_pred))
```

```
[1 0 0 1 1 1 1 1 1 1 0 0 1 0 0 1 0 0 1 1 0 1 1 0 1 1 1 1 1 0 0 0 0 0 1 0
 1 1 1 1 0 1 0 0 1 0 0 1 1 0 1 0 1 1 1 1 1 0 1 1 0 0 0 0 1 0 1 0 0 1 1 0 0
 0 0 1 0 0 0 0 0 0 1 0 0 1 0 1 1 1 1 1 0 0 1 1 1 0 1 1 1 1 1 1 1 0 1 1 0]
```

Saved successfully!



```
[[0 0]
 [ 0 62]]
```

	precision	recall	f1-score	support
0	1.00	0.95	0.97	58
1	0.95	1.00	0.98	62
accuracy			0.97	120
macro avg	0.98	0.97	0.97	120
weighted avg	0.98	0.97	0.97	120

```
# GradientBoostingClassifier:
from sklearn.ensemble import GradientBoostingClassifier
GradientBoost = GradientBoostingClassifier()
GradientBoost = GradientBoost.fit(X_train,y_train)
```

```
# Predictions:
y_pred = GradientBoost.predict(X_test)
print(y_pred)
# Performance:
print('Accuracy:', accuracy_score(y_test,y_pred))
print(confusion_matrix(y_test,y_pred))
print(classification_report(y_test,y_pred))
```

```
[1 0 0 1 1 1 1 1 1 1 0 0 1 0 0 1 0 0 1 1 0 1 1 0 1 1 1 1 1 0 0 0 0 0 1 0
 1 1 1 1 0 1 0 0 1 0 0 1 1 0 1 0 1 1 1 1 0 1 1 0 0 0 0 1 0 1 0 0 1 1 0 0
 0 0 1 0 0 0 0 0 0 1 0 0 1 0 1 1 1 1 1 0 0 1 1 1 0 1 1 1 1 1 1 1 0 1 1 0
 1 1 1 0 1 0 0 0 1]
```

Accuracy: 0.975

```
[[55  3]
 [ 0 62]]
```

	precision	recall	f1-score	support
0	1.00	0.95	0.97	58
1	0.95	1.00	0.98	62

0.97	120
0.97	120
0.97	120

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