```
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
from sklearn.model_selection import train_test_split,GridSearchCV
from sklearn.preprocessing import OneHotEncoder,MinMaxScaler,StandardScaler,LabelEncoder
from \ sklearn.compose \ import \ ColumnTransformer, make\_column\_selector
from \ sklearn. metrics \ import \ classification\_report, roc\_auc\_score, Confusion Matrix Display, f1\_score, accuracy\_score
{\tt from \ sklearn.exceptions \ import \ NotFittedError}
import tensorflow as tf
sns.set_style('darkgrid')
from sklearn.model_selection import (train_test_split,KFold,RandomizedSearchCV)
from scipy.stats import (randint,uniform)
from xgboost import XGBClassifier
from sklearn.metrics import (accuracy_score,confusion_matrix,classification_report)
df = pd.read_csv("/content/waterQuality.csv")
df.head()
         aluminium ammonia arsenic barium cadmium chloramine chromium copper flouride
      0
               1.65
                        9.08
                                 0.04
                                          2.85
                                                  0.007
                                                                0.35
                                                                          0.83
                                                                                   0.17
                                                                                             0.05
      1
               2.32
                       21.16
                                 0.01
                                          3.31
                                                  0.002
                                                                5.28
                                                                          0.68
                                                                                   0.66
                                                                                             0.90
      2
               1.01
                       14.02
                                 0.04
                                          0.58
                                                  0.008
                                                                4.24
                                                                          0.53
                                                                                   0.02
                                                                                             0.99
      3
               1.36
                       11.33
                                          2.96
                                                  0.001
                                                                7.23
                                                                          0.03
                                                                                   1.66
                                                                                             1.08
                                 0.04
```

df.isnull().sum()

4

0.92

5 rows × 21 columns

24.33

0.03

0.20

0.006

2.67

0.69

0.57

0.61

aluminium ammonia 0 arsenic 0 harium 9 cadmium 0 chloramine 0 chromium copper 0 flouride 0 bacteria viruses 0 lead 0 nitrates nitrites 0 mercury 0 perchlorate 0 radium 0 selenium 0 silver 0 uranium 0 is_safe 0 dtype: int64

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7999 entries, 0 to 7998
Data columns (total 21 columns):

#	Column	Non-i	Null Count	Dtype
0	aluminium	7999	non-null	float64
1	ammonia	7999	non-null	object
2	arsenic	7999	non-null	float64
3	barium	7999	non-null	float64
4	cadmium	7999	non-null	float64
5	chloramine	7999	non-null	float64
6	chromium	7999	non-null	float64
7	copper	7999	non-null	float64
8	flouride	7999	non-null	float64
9	bacteria	7999	non-null	float64
10	viruses	7999	non-null	float64
11	lead	7999	non-null	float64

```
12 nitrates
                 7999 non-null
                                 float64
13 nitrites
                 7999 non-null
                                float64
14 mercury
                 7999 non-null
                                 float64
15 perchlorate 7999 non-null
                                 float64
16 radium
                 7999 non-null
                                 float64
17 selenium
                 7999 non-null
                                 float64
18 silver
                 7999 non-null
                                 float64
                                                                       Loading...
19 uranium
                 7999 non-null
                                float64
                 7999 non-null
20 is_safe
                                object
dtypes: float64(19), object(2)
memory usage: 1.3+ MB
```

df.describe()

```
aluminium
                        arsenic
                                      barium
                                                   cadmium
                                                            chloramine
                                                                            chromium
count 7999.000000 7999.000000 7999.000000 7999.000000 7999.000000 7999.000000 7999.
mean
          0.666158
                       0.161445
                                     1.567715
                                                  0.042806
                                                               2.176831
                                                                             0.247226
                                                                                          0.
 std
          1.265145
                       0.252590
                                     1.216091
                                                  0.036049
                                                               2.567027
                                                                            0.270640
                                                                                          0.
min
          0.000000
                       0.000000
                                     0.000000
                                                  0.000000
                                                               0.000000
                                                                            0.000000
                                                                                          0.
                                                  0.008000
25%
          0.040000
                       0.030000
                                     0.560000
                                                               0.100000
                                                                             0.050000
                                                                                          0.
          0.070000
                       0.050000
                                                               0.530000
50%
                                     1.190000
                                                  0.040000
                                                                            0.090000
                                                                                          0.
75%
          0.280000
                       0.100000
                                     2.480000
                                                  0.070000
                                                               4.240000
                                                                             0.440000
                                                                                          1.
          5.050000
                       1.050000
                                     4 940000
                                                  0.130000
                                                               8.680000
                                                                            0.900000
                                                                                          2
max
```

```
target = 'is_safe'
df.loc[[7551, 7568,7890],'ammonia']
     7551
             #NUM!
     7568
             #NUM!
     7890
            #NUM!
     Name: ammonia, dtype: object
df.loc[[7551, 7568,7890],'is_safe']
     7551
             #NUM!
             #NUM!
     7568
             #NUM!
     7890
     Name: is_safe, dtype: object
df.drop([7551,7568,7890],axis=0,inplace=True)
df['ammonia']=df['ammonia'].astype('float64')
df['is_safe']=df['is_safe'].astype('int64')
num_col=[]
for col in df.select_dtypes(exclude='object'):
    if(col!=target):
       print(f"{col:30} ----> {df[col].nunique():10} unique values")
       num_col.append(col)
     aluminium
                                    ---->
                                                495 unique values
     ammonia
                                    --->
                                                2563 unique values
     arsenic
                                    --->
                                                107 unique values
                                    --->
     barium
                                                480 unique values
     cadmium
                                    --->
                                                 23 unique values
                                    --->
     chloramine
                                                812 unique values
                                    --->
                                                 91 unique values
     chromium
     copper
                                    --->
                                                201 unique values
     flouride
                                    --->
                                                151 unique values
     bacteria
                                    --->
                                                101 unique values
                                    --->
                                                 61 unique values
     viruses
     lead
                                    --->
                                                200 unique values
     nitrates
                                    --->
                                               1803 unique values
     nitrites
                                    --->
                                                280 unique values
                                    --->
     mercury
                                                 11 unique values
     perchlorate
                                    --->
                                                2999 unique values
                                    --->
                                                735 unique values
     radium
                                    ---->
     selenium
                                                 11 unique values
```

silver ----> 51 unique values uranium ----> 10 unique values

df.info()

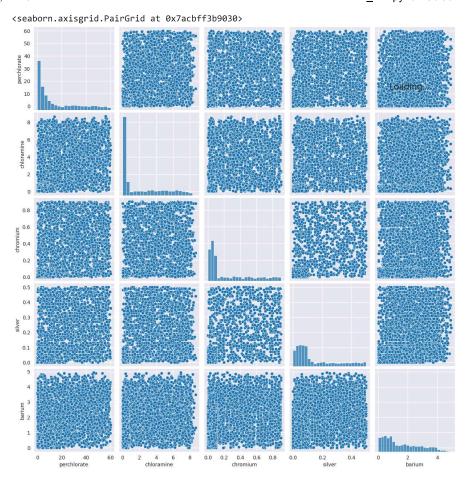
```
<class 'pandas.core.frame.DataFrame'>
Index: 7996 entries, 0 to 7998
Data columns (total 21 columns):
#
    Column
                  Non-Null Count Dtype
0
                  7996 non-null
     aluminium
                                  float64
 1
     ammonia
                  7996 non-null
                                   float64
 2
     arsenic
                  7996 non-null
                                  float64
 3
     barium
                  7996 non-null
                                  float64
     cadmium
                  7996 non-null
                                   float64
 5
     chloramine
                  7996 non-null
                                   float64
 6
     chromium
                  7996 non-null
                                   float64
                  7996 non-null
                                  float64
     copper
 8
     flouride
                  7996 non-null
                                   float64
 9
     bacteria
                  7996 non-null
                                   float64
 10
    viruses
                  7996 non-null
                                   float64
                  7996 non-null
 11 lead
                                  float64
 12
    nitrates
                  7996 non-null
                                   float64
    nitrites
                  7996 non-null
                                   float64
                  7996 non-null
 14
    mercury
                                   float64
 15
    perchlorate
                  7996 non-null
                                   float64
    radium
                  7996 non-null
                                   float64
                  7996 non-null
                                   float64
 17
    selenium
 18
    silver
                  7996 non-null
                                   float64
 19
    uranium
                  7996 non-null
                                   float64
20 is safe
                  7996 non-null
                                   int64
dtypes: float64(20), int64(1)
memory usage: 1.3 MB
```

Loading...

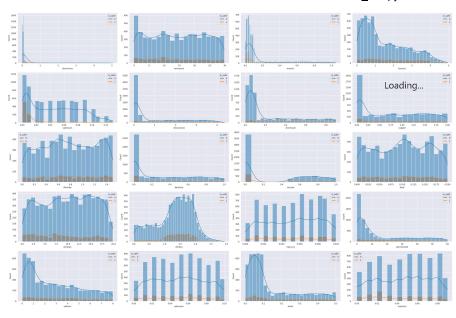
plt.figure(figsize=(20,10))
sns.heatmap(df.corr(), annot=True)
plt.show()

```
new_df = df[['perchlorate','chloramine','chromium','silver','barium']]
```

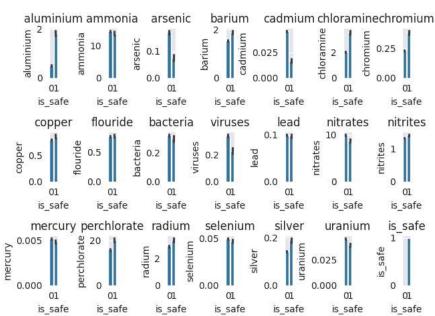
sns.pairplot(new_df)



```
plt.figure(figsize=(30,20))
j=1
for i,columns in enumerate(num_col):
    plt.subplot(5,4,j)
    sns.histplot(x=columns,data=df,kde=True,hue=target)
    j+=1
plt.tight_layout()
```







```
oh=OneHotEncoder(drop='first',handle_unknown='ignore')
ms,ss=MinMaxScaler(),StandardScaler()
ct=ColumnTransformer([
                                                                                 Loading...
    ('num_encoder',ss,make_column_selector(dtype_include='float64'))
],remainder='passthrough',verbose=True)
X=df.drop([target],axis=1)
y=df[[target]]
X.head(1)
         aluminium ammonia arsenic barium cadmium chloramine chromium copper flouride bacteria viruses
                                                                                                                    lead
                                                                                                                          nitrates nitrites me
               1.65
                        9.08
                                 0.04
                                         2.85
                                                 0.007
                                                              0.35
                                                                         0.83
                                                                                 0.17
                                                                                           0.05
                                                                                                       0.2
                                                                                                               0.0 0.054
                                                                                                                               16.08
                                                                                                                                          1.13
              Generate code with X
                                      View recommended plots
 Next steps:
y.head(1)
                    丽
      0
              Generate code with \, y \,
                                      View recommended plots
 Next steps:
X\_train, X\_test, y\_train, y\_test=train\_test\_split(X, y, test\_size=0.20, stratify=y, random\_state=42)
X_{train.shape,y_{train.shape}}
     ((6396, 20), (6396, 1))
X_test.shape,y_test.shape
     ((1600, 20), (1600, 1))
X_test.head(2)
            aluminium ammonia arsenic barium cadmium chloramine chromium copper flouri
      1264
                                    0.41
                                                                            0.06
                                                                                    0.28
                                                                                              0
                  3.90
                          22.64
                                            0.34
                                                    0.030
                                                                  7.94
                                    0.04
                                                    0.007
                                                                            0.86
                                                                                              0
      180
                  4 08
                          20 19
                                            3 34
                                                                  5 23
                                                                                    1.94
 Next steps:
              Generate code with X_test
                                           View recommended plots
y_test.head(2)
                       1264
                       ıl.
       180
                  0
                                           View recommended plots
 Next steps:
              Generate code with y_test
X_train, X_test, y_train, y_test = train_test_split(df.drop('is_safe', axis = 1),
                                                      df['is_safe'],
                                                      test_size = 0.2,
                                                      random_state = 0)
```

```
model = XGBClassifier()
param\_grid = {
    'n_estimators': randint(50, 251),
    'max_depth': randint(3, 15),
    'min_child_weight': randint(1, 11),
                                                                                  Loading...
    'gamma': uniform(0.0, 1.0),
}
kf = KFold(n_splits = 5, shuffle = True, random_state = 0)
search = RandomizedSearchCV(model,
                             param_grid,
                             scoring = 'accuracy',
                             cv = kf,
                             n_{iter} = 100,
                             refit = True,
                             n_{jobs} = -1)
search.fit(X_train, y_train)
     /usr/local/lib/python3.10/dist-packages/joblib/externals/loky/backend/fork_exec.py:38:
           RandomizedSearchCV
      ▶ estimator: XGBClassifier
search.best_params_
     {'gamma': 0.44137073152774653,
       'max_depth': 6,
      'min_child_weight': 1,
      'n_estimators': 154}
search.best_score_
     0.9710759137998437
print(f'Train Score : {accuracy_score(y_train, search.predict(X_train))}')
print(f'Test Score : {accuracy_score(y_test, search.predict(X_test))}')
     Train Score : 0.9984365228267668
     Test Score : 0.968125
{\tt confusion\_matrix}({\tt y\_test}, \ {\tt search.predict}({\tt X\_test}))
     array([[1407, 12],
            [ 39, 142]])
print(classification_report(y_test, search.predict(X_test)))
                   precision
                                recall f1-score
                                                    support
                0
                        0.97
                                   0.99
                                             0.98
                                                        1419
                        0.92
                                   0.78
                                             0.85
                                                         181
                                                        1600
         accuracy
                                             0.97
        macro avg
                        0.95
                                   0.89
                                             0.91
                                                        1600
                        0.97
                                             0.97
                                                        1600
     weighted avg
                                   0.97
```