## Importamos la librerias

```
In [41]:
          import pandas as pd
          from sklearn.model_selection import train_test_split
          from sklearn.preprocessing import StandardScaler
          from sklearn.linear model import LogisticRegression
          from sklearn.metrics import confusion matrix
          from sklearn.metrics import precision score
          from sklearn.metrics import accuracy_score
          df = pd.read csv('./water potability.csv')
In [42]:
          df=df.dropna()
          df.sample(10)
                                           Solids
                                                 Chloramines
Out[42]:
                     ph
                           Hardness
                                                                 Sulfate Conductivity
                                                                                      Organic_carbon Tri
          1110 4.736405 203.276420 28698.729965
                                                     5.271367 323.683683
                                                                           508.820341
                                                                                           14.063140
          2243 6.014540 205.482940 23268.243081
                                                     9.035769 283.340368
                                                                           371.403475
                                                                                           16.863668
          2694 3.676845 173.677702 22581.224488
                                                    12.653362 289.126846
                                                                           599.729246
                                                                                            8.386407
          3128 4.959853 215.854869
                                                     6.954231 379.504731
                                                                                           14.326638
                                     9887.830755
                                                                           527.479694
           212 5.304143 160.968358
                                    25299.511557
                                                     6.286043 300.500689
                                                                           497.790391
                                                                                           15.013030
          1453 8.034012 197.609392
                                    25204.674082
                                                     6.501097 330.552173
                                                                           490.821500
                                                                                           15.107289
          1434 7.322502 196.104298
                                                     6.661157 297.841381
                                    19534.421648
                                                                           418.075511
                                                                                           17.161768
          2084 7.094831 139.906866
                                    18975.643495
                                                     7.571951 423.045974
                                                                           528.266523
                                                                                           18.556177
          2167 8.180755 178.079638
                                                     7.574082 391.918229
                                                                           353.507014
                                                                                           19.749408
                                    25623.867503
           523 8.437876 183.106625 12856.928695
                                                     7.400019 361.779682
                                                                           587.022609
                                                                                           15.926921
          Selecciono las variables
          parametros = df.drop(columns="Potability")
In [43]:
          objetivo = df.Potability
          Implementacion del model
          Preparamos los datos de prueba
          X_train, X_test, y_train,y_test =train_test_split(parametros,objetivo,test_size=0.5)
In [44]:
          Escalamos los datos
          escalar = StandardScaler()
In [45]:
          X_train = escalar.fit_transform(X_train)
In [46]:
          X test = escalar.transform(X test)
```

Defino el modelo

```
In [47]:
         model = LogisticRegression()
         Entreno el modelo
         model.fit(X_train,y_train)
In [48]:
         LogisticRegression()
Out[48]:
         Realizo una prediccion
         y_pred = model.predict(X_test)
In [49]:
         Verifico la prediccion
         matriz = confusion matrix(y test,y pred)
In [50]:
          print("Matriz de Confusion", matriz)
         Matriz de Confusion [[590 30]
           [365 21]]
         Precision del Modelo
         precision = precision_score(y_test,y_pred)
In [51]:
          print("Precision del modelo", precision)
         Precision del modelo 0.4117647058823529
         Exactitud del Modelo
         exactitud = accuracy_score(y_test,y_pred)
In [52]:
          print("Exactitud del modelo: ",exactitud)
         Exactitud del modelo: 0.6073558648111332
 In [ ]:
```