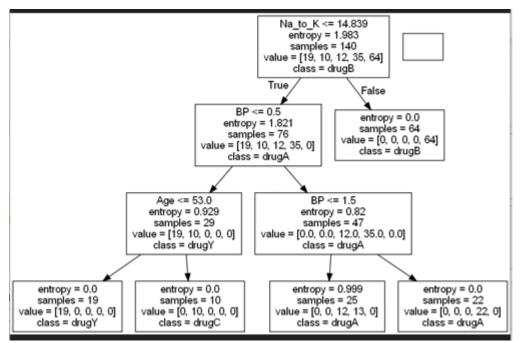
PRACTICA 23

```
In [2]: import pandas as pd
            import numpy as np
            import matplotlib.pyplot as plt
            from sklearn.tree import DecisionTreeClassifier
            from sklearn import metrics
            from sklearn.model_selection import train_test_split
            import warnings
            warnings.filterwarnings('ignore')
 In [3]: drugs = pd.read_csv('C:/Users/Isaac/Desktop/IHD/EBAC DT/M23 DS/drugs.csv')
            drugs
 Out[3]:
                                    BP Cholesterol Na_to_K Drug
                   Age Sex
                                  HIGH
                                                        25.355 drugY
                           F
                                               HIGH
               0
                    23
                1
                    47
                                  LOW
                                               HIGH
                                                        13.093 drugC
                          М
               2
                    47
                                  LOW
                                               HIGH
                                                         10.114 drugC
                           М
                           F NORMAL
                                               HIGH
               3
                    28
                                                         7.798 drugX
                    61
                           F
                                  LOW
                                               HIGH
                                                        18.043 drugY
In [4]: # defininimos caracteristicas para los pronosticos
        feature_cols = ['Age','Sex','BP','Cholesterol','Na_to_K']
       # definimos valores para 'x' y 'y'
       X = drugs[feature_cols].values
       y = drugs.Drug
        Convertir las variables predictoras cualitativas de esta base a una escala numérica mediante la instrucción "preprocessing.LabelEncoder()
In [5]: from sklearn import preprocessing
       cod_sex = preprocessing.LabelEncoder()
cod_sex.fit(['F','M'])
X[:,1] = cod_sex.transform(X[:,1])
       cod_bp = preprocessing.LabelEncoder()
       cod_bp.fit(['HIGH','LOW','NORMAL'])
X[:,2] = cod_bp.transform(X[:,2])
       cod_Chol = preprocessing.LabelEncoder()
cod_Chol.fit(['HIGH','NORMAL'])
X[:,3] = cod_Chol.transform(X[:,3])
                                                                                                           N
In [6]: # creamos grupos de entrenamiento y prueba
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3, random_state = 1)
In [7]: # objeto de calsificacion del Arbol de Decision
         clf = DecisionTreeClassifier()
         # entrenamos el modelo
         clf.fit(X_train, y_train)
Out[7]:
          DecisionTreeClassifier()
```

```
In [8]: # realizamos predicciones
                                  clf_pred = clf.predict(X_test)
  In [10]: # Matriz de Conficion
                                  from sklearn.metrics import confusion matrix
                                  cm = confusion_matrix(y_test, clf_pred)
                                  cm
  Out[10]: array([[ 4,
                                                                              0, 0, 0,
                                                                                                                       0],
                                                          [2, 4, 0, 0, 0],
                                                          [0, 0, 4, 0, 0],
                                                          [0, 0, 0, 19, 0],
                                                          [ 0, 0, 0, 0, 27]], dtype=int64)
                                      ESTADISTICAS DE DESEMPEÑO
       In [11]: from sklearn.metrics import classification report
                                      print(classification_report(y_test, clf_pred))
                                                                                     precision
                                                                                                                               recall f1-score
                                                                                                                                                                                               support
                                                              drugA
                                                                                                      0.67
                                                                                                                                      1.00
                                                                                                                                                                        0.80
                                                                                                                                                                                                                  4
                                                             drugB
                                                                                                     1.00
                                                                                                                                      0.67
                                                                                                                                                                       0.80
                                                                                                                                                                                                                  6
                                                                                                                                                                      1.00
                                                              drugC
                                                                                                     1.00
                                                                                                                                      1.00
                                                                                                                                                                                                                  4
                                                                                                                                                                   1.00
                                                              drugX
                                                                                                     1.00
                                                                                                                                      1.00
                                                                                                                                                                                                               19
                                                              drugY
                                                                                                     1.00
                                                                                                                                      1.00
                                                                                                                                                                       1.00
                                                                                                                                                                                                               27
                                                                                                                                                                       0.97
                                                   accuracy
                                                                                                                                                                                                               60
                                                                                                     0.93
                                                                                                                                                                       0.92
                                                                                                                                                                                                               60
                                                macro avg
                                                                                                                                      0.93
                                      weighted avg
                                                                                                     0.98
                                                                                                                                      0.97
                                                                                                                                                                       0.97
                                                                                                                                                                                                               60
                 ARBOL DE DECISION (REGLAS)
In [13]: dot_data = tree.export_graphviz(clf, out_file = None, feature_names = feature_cols, class_names = ['drugY', 'drugX', 'dr
                 # creamos la grafica de Arbol
                 graph = pydotplus.graph_from_dot_data(dot_data)
In [14]: # creacion del Arbol en formato png
                 graph.write_png('drug.png')
image = Image.open('drug.png')
image.show()
```



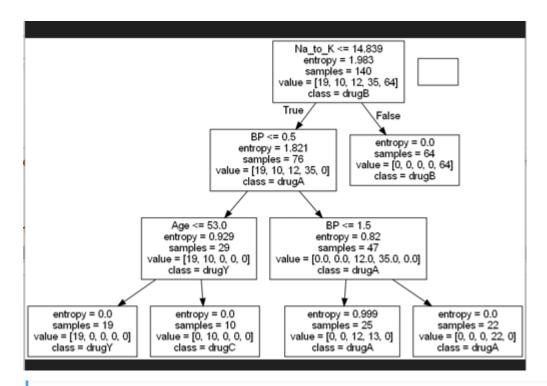
utilizando entropy

```
In [16]: clf = DecisionTreeClassifier(criterion = 'entropy', max_depth = 3)
    clf.fit(X_train, y_train)
    clf_pred = clf.predict(X_test)
```

In [18]: from sklearn.metrics import classification_report
 print(classification_report(y_test, clf_pred))

	precision	recall	f1-score	support
drugA	0.67	1.00	0.80	4
drugB	1.00	0.67	0.80	6
drugC	0.00	0.00	0.00	4
drugX	0.83	1.00	0.90	19
drugY	1.00	1.00	1.00	27
accuracy			0.90	60
macro avg	0.70	0.73	0.70	60
weighted avg	0.86	0.90	0.87	60

```
In [19]: dot_data = tree.export_graphviz(clf, out_file = None, feature_names = feature_cols, class_names = ['drugY','drugC','drugX','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drugA','drug
```



Age Sex BP

¿Qué medicamento recomendaría utilizar para un paciente con los siguientes datos?

Cholesterol Na_to_K



* El medicamento que se recomendaria al paciente con los datos dados, sería Drug Y