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Question 4: Kaggle Dataset

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
df = pd.read csv('/content/Heart.csv')
df.head()
{"summary":"{\n \"name\": \"df\",\n \"rows\": 303,\n \"fields\": [\
n {\n \"column\": \"Unnamed: 0\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 87,\n \"min\": 1,\n
\"max\": 303,\n \"num_unique_values\": 303,\n \"samples\": [\n 180,\n 229,\n 112\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n },\n {\n \"column\": \"Age\",\n \"properties\": {\
\"num_unique_values\": 2,\n \"samples\": [\n
1\n ],\n \"semantic_type\": \"\",\n
                                                                                0.\n
\"description\": \"\"\n }\n },\n {\n \"column\":
\"ChestPain\",\n \"properties\": {\n \"dtype\":
\"category\",\n \"num_unique_values\": 4,\n \"samples\":
[\n \"asymptomatic\",\n \"nontypical\"\n ],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"RestBP\",\n \"properties\":
{\n \"dtype\": \"number\",\n \"std\": 17,\n \\"min\": 94,\n \"max\": 200,\n \"num_unique_values\":
\"num_unique_values\": 152,\n \"samples\": [\n
187\n ],\n \"semantic_type\": \"\",\n
                                                                                   321,\n
\"description\":\"\n \\n \\n\\"column\":\\"Fbs\\",\n \\"properties\\": \\n \\"dtype\\":\\"number\\",\n \\"std\\": 0,\n \\"min\\": 0,\n \\"max\\": 1,\n
\"num_unique_values\": 2,\n \"samples\": [\n
1\n ],\n \"semantic_type\": \"\",\n
                                                                                0, n
```

```
\"RestECG\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 0,\n \"min\": 0,\n \"max\": 2,\n \"num_unique_values\": 3,\n \"samples\": [\n 2,\n
\"num_unique_values\": 2,\n \"samples\": [\n 1,\n 0\n ],\n \"semantic_type\": \"\",\n \"dtype\": \"number\",\n \"std\": 1.161075022068634,\n \"min\": 0.0,\n
\"max\": 6.2,\n \"num_unique_values\": 40,\n \"samples\": [\n 2.4,\n 0.2\n
                                                     0.2\n
\"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"Slope\",\n \"properties\": {\
n \"dtype\": \"number\",\n \"std\": 0,\n \
1,\n \"max\": 3,\n \"num_unique_values\": 3,\n \
\"samples\": [\n 3,\n 2\n ],\n \
\"semantic_type\": \"\",\n \"description\": \"\"\n \
\"semantic_type\": \"\",\n \"description\": \"\"\n \\"semantic_type\": \"\"\n \"\"
           \"dtype\": \"number\",\n \"std\": 0,\n \"min\":
n },\n {\n \"column\": \"Ca\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 0.9374383177242157,\n \"min\": 0.0,\n \"max\": 3.0,\n \"num_unique_values\": 4,\n \"samples\": [\n 3.0,\n 1.0\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
n}","type":"dataframe","variable name":"df"}
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 303 entries, 0 to 302
Data columns (total 15 columns):
                      Non-Null Count Dtype
      Column
 0
       Unnamed: 0 303 non-null
                                          int64
 1
      Age 303 non-null int64
```

```
2
     Sex
                  303 non-null
                                   int64
 3
     ChestPain
                  303 non-null
                                   object
 4
     RestBP
                  303 non-null
                                   int64
5
     Chol
                  303 non-null
                                   int64
 6
     Fbs
                  303 non-null
                                   int64
 7
     RestECG
                  303 non-null
                                   int64
 8
                  303 non-null
     MaxHR
                                   int64
 9
     ExAng
                  303 non-null
                                   int64
 10
    Oldpeak
                  303 non-null
                                   float64
 11
     Slope
                  303 non-null
                                   int64
                  299 non-null
                                   float64
 12
     Ca
13
     Thal
                  301 non-null
                                   object
14
     AHD
                  303 non-null
                                   object
dtypes: float64(2), int64(10), object(3)
memory usage: 35.6+ KB
df.shape
(303, 15)
df.isnull().sum()
Unnamed: 0
               0
Age
Sex
               0
ChestPain
               0
RestBP
               0
Chol
               0
               0
Fbs
RestECG
               0
               0
MaxHR
ExAng
               0
Oldpeak
               0
Slope
               0
Ca
               4
               2
Thal
AHD
               0
dtype: int64
df.dtypes
Unnamed: 0
                 int64
                 int64
Age
Sex
                 int64
ChestPain
                object
RestBP
                 int64
Chol
                 int64
Fbs
                 int64
RestECG
                 int64
MaxHR
                 int64
ExAng
                 int64
```

```
Oldpeak
              float64
Slope
               int64
Ca
              float64
Thal
               object
AHD
              object
dtype: object
for col in df.columns:
    zero count = (df[col] == 0).sum()
    print(f"Number of zeros in {col}: {zero count}")
Number of zeros in Unnamed: 0: 0
Number of zeros in Age: 0
Number of zeros in Sex: 97
Number of zeros in ChestPain: 0
Number of zeros in RestBP: 0
Number of zeros in Chol: 0
Number of zeros in Fbs: 258
Number of zeros in RestECG: 151
Number of zeros in MaxHR: 0
Number of zeros in ExAng: 204
Number of zeros in Oldpeak: 99
Number of zeros in Slope: 0
Number of zeros in Ca: 176
Number of zeros in Thal: 0
Number of zeros in AHD: 0
df['Age'].mean()
54.43894389438944
df.columns
Index(['Unnamed: 0', 'Age', 'Sex', 'ChestPain', 'RestBP', 'Chol',
'Fbs',
       'RestECG', 'MaxHR', 'ExAng', 'Oldpeak', 'Slope', 'Ca', 'Thal',
'AHD'],
     dtvpe='object')
#Now extract only Age, Sex, ChestPain, RestBP, Chol.
df.drop(columns=['Unnamed: 0', 'Fbs','RestECG', 'MaxHR', 'ExAng',
'Oldpeak', 'Slope', 'Ca', 'Thal'], inplace=True)
df.head()
{"summary":"{\n \"name\": \"df\",\n \"rows\": 303,\n \"fields\": [\
n {\n \"column\": \"Age\",\n \"properties\": {\n
                               \"std\": 9,\n \"min\": 29,\n
\"dtype\": \"number\",\n
\"max\": 77,\n \"num_unique_values\": 41,\n \"samples\": [\n 61,\n 64,\n 44\n ],\n
\"semantic_type\": \"\",\n
                                \"description\": \"\"\n
                   \"column\": \"Sex\",\n \"properties\": {\n
     },\n {\n
```

```
\"dtype\": \"number\",\n \"std\": 0,\n \"min\": 0,\n \"max\": 1,\n \"num_unique_values\": 2,\n \"samples\": [\n 0,\n 1\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n {\n
\"column\": \"ChestPain\",\n\"properties\": {\n\"dtype\": \"category\",\n\"num_unique_values\": 4,\n
\"samples\": [\n \"asymptomatic\",\n \"nontypical\"\
n ],\n \"semantic_type\": \"\",\n
\"num_unique_values\": 50,\n \"samples\": [\n 124,\n 192\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n }\n {\n \"column\": \"Chol\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 51,\n \"min\": 126,\n \"max\": 564,\n
\"num_unique_values\": 152,\n \"samples\": [\n 321,\n 187\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n \"num_unique_values\": {\n \"dtype\": \"category\",\n \"num_unique_values\": 2,\n \"samples\": [\n \"Yes\",\n \"No\"\n ],\n \"semantic_type\": \"\",\n \"dtype\": \"\",\n \"ho\"\n ],\n \"semantic_type\": \"\",\n \"ho\"\n ]\" \"semantic_type\": \"\",\n \"ho\"\n ]\" \"type\": \"\",\n \\"ho\"\n ]\" \"ho\"\"\n ]\" \"ho\"\"\n ]\" \"ho\"\"\n \\"ho\"\"\n \\"ho\"\"\"\n \\"ho\"\"\n \\"\n \\"ho\"\"\n \\"ho\"\n \\"ho\"\"\n \\"ho\"\"\n \\"ho\"\"\n \\"ho\"\"\n \\"ho\"\"\n \\"ho\"\"\n \\"ho\"\"\n \\"ho\"\"\n \\"ho\"\n \\"ho\"\n \\"ho\"\"\n \\"ho\"\"\n \\"ho\"\"\n \\"ho\"\n \\"ho
 n}","type":"dataframe","variable_name":"df"}
 x = df.drop(columns=['AHD'])
 y= df['AHD']
 display(x.head())
 display(y.head())
 {"summary":"{\n \"name\": \"display(y\",\n \"rows\": 5,\n
\"fields\": [\n {\n \"column\": \"Age\",\n
\"properties\": {\n \"dtype\": \"number\",\n \"std\":
[\n \"asymptomatic\",\n \"nontypical\"\n ],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"RestBP\",\n \"properties\":
{\n \"dtype\": \"number\",\n \"std\": 15,\n \\"min\": 120,\n \"max\": 160,\n \"num_unique_values\":
 4,\n \"samples\": [\n 160,\n
                                                                                                                                                      130\n ],\
```

```
\"semantic_type\": \"\",\n
                                          \"description\": \"\"\n
             {\n \"column\": \"Chol\",\n \"properties\":
      },\n
}\n
          \"dtype\": \"number\",\n
{\n
                                          \"std\": 30,\n
\"min\": 204,\n
                      \"max\": 286,\n
                                             \"num unique values\":
                              286,\n
5,\n
           \"samples\": [\n
                                                     204\n
        \"semantic_type\": \"\",\n
                                          \"description\": \"\"\n
n
      }\n ]\n}","type":"dataframe"}
}\n
0
     No
1
    Yes
2
    Yes
3
     No
4
     No
Name: AHD, dtype: object
y.replace({'Yes': 1, 'No': 0}, inplace=True)
display(y.head())
<ipython-input-12-a8acbb79871c>:1: FutureWarning: Downcasting behavior
in `replace` is deprecated and will be removed in a future version. To
retain the old behavior, explicitly call
`result.infer objects(copy=False)`. To opt-in to the future behavior,
set `pd.set option('future.no silent downcasting', True)`
 y.replace(\{'Yes': 1, 'No': \overline{0}\}, inplace=True)
0
1
     1
2
     1
3
    0
4
Name: AHD, dtype: int64
from sklearn.model selection import train test split
x train, x test, y train, y test = train test split(x,y,
test size=0.25, random state=90)
print(x train.shape)
display(x train.head())
print(x_test.shape)
display(x test.head())
print(y train.shape)
display(y_train.head())
print(y test.shape)
display(y test.head)
(227, 5)
{"summary":"{\n \"name\": \"display(y_test\",\n \"rows\": 5,\n
\"fields\": [\n {\n
                          \"column\": \"Age\",\n
                        \"dtype\": \"number\",\n
\"properties\": {\n
                                                          \"std\":
           \"min\": 40,\n
                                 \"max\": 62,\n
8,\n
                                \"samples\": [\n
\"num unique values\": 5,\n
                                                            59,\n
```

```
\"Sex\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 0,\n \"min\": 0,\n \"max\": 1,\n
\"num_unique_values\": 2,\n \"samples\": [\n
0\n ],\n \"semantic_type\": \"\",\n
                                                                    1.\n
\"description\": \"\"\n }\n }\n {\n \"column\":
\"ChestPain\",\n \"properties\": {\n \"dtype\":
\"string\",\n \"num_unique_values\": 3,\n \"samples
                       \"num unique values\": 3,\n \"samples\":
[\n \"asymptomatic\",\n \"typical\"\n ],\n
\"semantic_type\": \"\",\n \"description\": \"\"\n }\
n },\n {\n \"column\": \"RestBP\",\n \"properties\":
{\n \"dtype\": \"number\",\n \"std\": 19,\n \"min\": 132,\n \"max\": 178,\n \"num_unique_values\":
4,\n \"samples\": [\n 178,\n 140\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
}\n }\n ]\n}","\undergreen";"dataframe"}
(76, 5)
{"summary":"{\n \"name\": \"display(y test\",\n \"rows\": 5,\n
3,\n \"min\": 53,\n \"max\": 61,\n
\"num_unique_values\": 4,\n \"samples\": [\n 53,\n
59,\n 56\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n {\n \"column\":
\"Sex\",\n \"properties\": {\n \"dtype\": \"number\",\n
\"std\": 0,\n \"min\": 1,\n \"max\": 1,\n
\"num unique values\": 1\n \"samples\": [\n 1\n
\"num_unique_values\": 1,\n \"samples\": [\n
\"num_unique_values\": 4,\n \"samples\": [\n
                                                                     123\n
```

```
\"semantic_type\": \"\",\n \"description\": \"\"\n
                                                                 }\
     }\n ]\n}","type":"dataframe"}
(227,)
209
       1
183
       0
       0
165
41
       0
291
       0
Name: AHD, dtype: int64
(76,)
<bound method NDFrame.head of 288 0</pre>
264
       1
96
       1
       0
219
55
       1
79
       1
217
       0
287
       0
188
       1
Name: AHD, Length: 76, dtype: int64>
x train['ChestPain'].unique()
array(['asymptomatic', 'typical', 'nontypical', 'nonanginal'],
      dtype=object)
from sklearn.preprocessing import OrdinalEncoder
ordinal encoder = OrdinalEncoder(categories=[['typical',
'asymptomatic', 'nonanginal', 'nontypical']])
x train['ChestPain'] =
ordinal encoder.fit transform(x train[['ChestPain']])
x_test['ChestPain'] = ordinal_encoder.transform(x_test[['ChestPain']])
print(x train.head())
print(x test.head())
     Age Sex ChestPain
                          RestBP
                                   Chol
209
      62
            0
                     1.0
                              150
                                    244
183
      59
            1
                     0.0
                             178
                                    270
            1
                     1.0
                             132
                                   207
165
      57
41
      40
            1
                     0.0
                             140
                                    199
291
      55
                     3.0
                                    342
            0
                             132
     Age Sex ChestPain RestBP
                                   Chol
288
      56
                                    221
            1
                     3.0
                             130
            1
223
      53
                     1.0
                              123
                                    282
```

```
264
      61
            1
                    1.0
                                  166
                            138
                                  239
96
      59
            1
                    1.0
                            110
219
      59
            1
                    1.0
                            138
                                  271
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
x train[['Age', 'RestBP', 'Chol']] =
scaler.fit transform(x train[['Age', 'RestBP', 'Chol']])
x test[['Age', 'RestBP', 'Chol']] = scaler.transform(x test[['Age',
'RestBP', 'Chol']])
print(x train.head())
print(x test.head())
          Age Sex ChestPain
                                RestBP
                                            Chol
209
     0.834084
                0
                          1.0
                              1.014683 -0.025148
                1
183
    0.508203
                         0.0 2.627015 0.469594
165
    0.290948
                 1
                          1.0 -0.021816 -0.729203
41 -1.555713
                1
                         0.0 0.438850 -0.881431
291
    0.073694
                          3.0 -0.021816 1.839647
                 0
              Sex ChestPain
          Age
                                RestBP
                                            Chol
288
    0.182321
                         3.0 -0.136982 -0.462804
                1
                1
223 -0.143560
                          1.0 -0.540065
                                        0.697936
                1
264
    0.725457
                          1.0 0.323684 -1.509372
96
     0.508203
                 1
                          1.0 -1.288648 -0.120290
219
                1
    0.508203
                         1.0 0.323684 0.488622
from sklearn.linear model import LogisticRegression
model = LogisticRegression()
model.fit(x train, y train)
LogisticRegression()
y pred = model.predict(x test)
y pred
array([0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1,
0,
       1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0,
1,
       0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0,
0,
       1, 0, 0, 0, 1, 1, 1, 0, 0, 0])
pd.DataFrame({'y_test':y_test, 'y_predict':y_pred})
{"summary":"{\n \"name\": \"pd\",\n \"rows\": 76,\n \"fields\": [\n
       \"column\": \"y test\",\n \"properties\": {\n
{\n
\"dtype\": \"number\",\n
                               \"std\": 0,\n
                                                 \"min\": 0,\n
\"max\": 1,\n \"num unique values\": 2,\n
                                                      \"samples\":
```

```
1,\n
                            0\n
                                                     \"semantic type\":
\lceil \backslash n \rceil
                                       1,\n
\"\",\n
              \"description\": \"\"\n
                                              }\n
                                                     },\n
                                                             {\n
\"column\": \"y_predict\",\n \"properties\": {\n
\"dtype\": \"number\",\n
                                                        \"min\": 0,\n
                                  \"std\": 0,\n
\"max\": 1,\n
                \"num unique values\": 2,\n
                                                       \"samples\":
                                                     \"semantic type\":
[\n
                            0\n
                                       ],\n
               \"description\": \"\"\n
\"\",\n
                                                     }\n ]\
n}","type":"dataframe"}
from sklearn.metrics import accuracy score, precision score,
recall_score, fl_score, confusion_matrix, classification report
print('Accuracy Score: ', accuracy_score(y_test, y_pred))
print('Precision Score: ', precision_score(y_test, y_pred))
print('Recall Score: ', recall_score(y_test, y_pred))
print('F1 Score: ', f1 score(y test, y pred))
print('Confusion Matrix:\n', confusion matrix(y test, y pred))
print('Classification Report:\n', classification report(y test,
y pred))
Accuracy Score:
                 0.75
Precision Score: 0.7428571428571429
Recall Score: 0.722222222222222
F1 Score: 0.7323943661971831
Confusion Matrix:
  [[31 9]
 [10 26]]
Classification Report:
                 precision recall f1-score
                                                  support
                              0.78
                                                      40
           0
                    0.76
                                         0.77
           1
                    0.74
                              0.72
                                         0.73
                                                      36
                                         0.75
                                                      76
    accuracy
   macro avq
                    0.75
                              0.75
                                         0.75
                                                      76
weighted avg
                    0.75
                              0.75
                                         0.75
                                                      76
sns.heatmap(confusion matrix(y test, y pred), annot=True, fmt='d')
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.show()
```

