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import pandas as pd
from sklearn.linear model import LogisticRegression
from sklearn.semi supervised import SelfTrainingClassifier
from sklearn.preprocessing import LabelEncoder
import numpy as np
import seaborn as sns
df = pd.read csv("/content/Titanic Data (2).csv")
df = df[["Survived", "Pclass", "Sex", "Age", "Fare"]].dropna()
le = LabelEncoder()
df["Sex"] = le.fit transform(df["Sex"])
np.random.seed(42)
mask = np.random.rand(len(df)) < 0.5
df.loc[mask, "Survived"] = -1
X = df.drop("Survived", axis=1)
y = df["Survived"]
base model = LogisticRegression()
self training model = SelfTrainingClassifier(base model,
criterion='k best')
self training model.fit(X, y)
SelfTrainingClassifier(criterion='k best',
estimator=LogisticRegression())
SelfTrainingClassifier(criterion='k best',
estimator=LogisticRegression())
SelfTrainingClassifier(criterion='k best',
estimator=LogisticRegression())
df["Predicted Label"] = self training model.predict(X)
print(df.head(20))
   Survived Pclass Sex Age
                                          Predicted Label
                                    Fare
                  3
                       1 34.5
0
          - 1
                                  7.8292
                                                        0
1
           1
                   3
                        0 47.0
                                  7.0000
                                                        1
2
                   2
                       1 62.0
                                                        0
           0
                                  9.6875
3
                  3
           0
                       1 27.0 8.6625
                                                        0
4
          - 1
                   3
                        0 22.0 12.2875
                                                        1
5
                  3
          - 1
                       1 14.0 9.2250
                                                        0
6
          - 1
                   3
                       0 30.0 7.6292
                                                        1
                  2
7
           0
                      1 26.0 29.0000
                                                        0
8
                   3
          1
                      0 18.0 7.2292
                                                        1
9
           0
                   3
                       1 21.0 24.1500
                                                        0
11
          - 1
                        1 46.0 26.0000
                                                        0
```

20 -1 1 1 55.0 59.4000 0	12 13 14 15 16 17 18 19 20	1 0 -1 -1 -1 -1 -1	1 2 1 2 2 3 3 3	0 1 0 0 1 1 0 0	23.0 63.0 47.0 24.0 35.0 21.0 27.0 45.0 55.0	82.2667 26.0000 61.1750 27.7208 12.3500 7.2250 7.9250 7.2250 59.4000	1 0 1 1 0 0 1 1
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