| [54]: [55]: F | df.head() RID age income student credit_rating buys_computer 1 2 youth high no fair no 2 3 middle_aged high no fair yes 3 4 senior medium no fair yes 4 5 senior low yes fair yes |
|-----------------------------|---|
| [55]: F | 1 2 youth high no excellent no 2 3 middle_aged high no fair yes 3 4 senior medium no fair yes |
| [55]: F 6 2 5 6 | Preprocessing the data |
| | ##Preprocessing df.isnull().sum() RID 0 age 0 income 0 student 0 credit_rating 0 buys_computer 0 dtype: int64 |
| [56]: | RID age income student credit_rating buys_computer 1 1 2 youth high no excellent no excellent no |
| | 2 3 middle_aged high no fair yes 3 4 senior medium no fair yes 4 5 senior low yes fair yes 5 6 senior low yes excellent no 6 7 middle_aged low yes excellent yes 7 8 youth medium no fair no 8 9 youth low yes fair yes 9 10 senior medium yes fair yes 10 11 youth medium yes excellent yes |
| : | 11 12 middle_aged medium no excellent yes 12 13 middle_aged high yes fair yes 13 14 senior medium no excellent no import sklearn |
| [58]: | <pre>from sklearn.preprocessing import LabelEncoder l_age=LabelEncoder() l_income=LabelEncoder() l_student=LabelEncoder() l_credict_rating=LabelEncoder()</pre> |
| [59]: | <pre>l_buys_computer=LabelEncoder() df['ag']=l_age.fit_transform(df['age']) df['inco']=l_income.fit_transform(df['income']) df['stu']=l_student.fit_transform(df['student'])</pre> |
| 5007 | <pre>df['cr']=l_credict_rating.fit_transform(df['credit_rating']) df['bc']=l_buys_computer.fit_transform(df['buys_computer']) df RID age income student credit_rating buys_computer ag inco stu cr bc</pre> |
| | NB age micone stadent creating buys_compate ag micone sta t bc 0 1 youth high no fair no 2 0 0 0 0 2 3 middle_aged high no fair yes 0 0 0 1 1 3 4 senior medium no fair yes 1 2 0 1 1 4 5 senior low yes fair yes 1 |
| : | 8 9 youth low yes fair yes 2 1 1 1 9 10 senior medium yes fair yes 1 2 1 1 1 10 11 youth medium yes excellent yes 2 2 1 0 1 11 12 middle_aged medium no excellent yes 0 2 0 0 1 |
| | 12 13 middle_aged high yes fair yes 0 0 1 1 1 13 14 senior medium no excellent no 1 2 0 0 0 df1=df.drop(['RID','age','income','student','credit_rating','buys_computer'],axis='columns') |
| [62]: | dfl ag inco stu cr bc 0 2 0 0 1 0 |
| | 1 2 0 0 0 0 0 1 1 3 3 1 2 0 1 1 1 4 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 |
| | feature_names=['ag', 'inco', 'stu', 'cr'] x=df[['ag', 'inco', 'stu', 'cr']] y=df['bc'] Training and Testing data |
| [65]: | <pre>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=42) from sklearn.tree import DecisionTreeClassifier DTC=DecisionTreeClassifier()</pre> |
| [65]: [| DTC=DecisionTreeClassifier() DTC.fit(x_train, y_train) DecisionTreeClassifier() df1.corr() |
| [07]. | ag inco stu cr bc ag 1.000000 0.092036 -8.944272e-02 7.745967e-02 -0.493333 inco 0.092036 1.000000 0.000000e+00 -1.980295e-01 0.127827 stu -0.089443 0.000000 1.000000e+00 -6.409876e-17 0.447214 cr 0.077460 -0.198030 -6.409876e-17 1.000000e+00 0.258199 bc -0.493333 0.127827 4.472136e-01 2.581989e-01 1.000000 |
| [67]: i | dfl·cov() rag inco stu cr bc ag 0.686813 0.065934 0.038462 0.038462 0.032967 0.203297 inco 0.038462 0.00000 0.269231 0.00000 0.115385 rag 0.032967 -0.087912 0.00000 0.263736 0.065934 bc 0.203297 0.054945 0.115385 0.065934 0.247253 |
| [68]: | <pre>sns.heatmap(df1.corr(), annot=True) </pre> <pre></pre> |
| [69]: | g - 0.49 |
| | - 18 |
| [70]: | Exploratory Data Analysis sns.pairplot(df1) <seaborn.axisgrid.pairgrid 0x280e631ac70="" at=""></seaborn.axisgrid.pairgrid> |
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| | 0.4 |
| | <pre>import six from six import StringIO from sklearn.tree import export_graphviz import pydotplus import graphviz from sklearn import tree</pre> |
| [72]: | <pre>dot_data_=StringIO() dot_data=tree.export_graphviz(DTC,</pre> |
| [72]. | $ \begin{array}{c} \text{graph=graphviz.Source(dot_data)} \\ \\ \text{graph} \\ \\ \\ \text{stu} <= 0.5 \\ \\ \text{gini} = 0.48 \\ \\ \text{samples} = 10 \\ \\ \text{value} = [4, 6] \\ \\ \text{class} = 1 \\ \\ \\ \text{True} \\ \\ \text{False} \\ \\ \text{samples} = 5 \\ \\ \\ \text{gini} = 0.48 \\ \\ \text{samples} = 5 \\ \\ \\ \text{samples} = 5 \\ \\ \end{array} $ |
| | |
| | $\begin{array}{c} \text{samples} = 1 \\ \text{value} = [1, 0] \\ \text{class} = 0 \end{array}$ $\begin{array}{c} \text{samples} = 2 \\ \text{value} = [0, 1] \\ \text{class} = 1 \end{array}$ $\begin{array}{c} \text{gini} = 0.0 \\ \text{samples} = 1 \\ \text{value} = [0, 1] \end{array}$ $\begin{array}{c} \text{gini} = 0.0 \\ \text{samples} = 1 \\ \text{value} = [0, 1] \end{array}$ $\begin{array}{c} \text{value} = [0, 1] \\ \text{value} = [0, 1] \end{array}$ |
| | <pre>class = 1</pre> |
| [75]: | <pre>from sklearn.metrics import classification_report classification_report(y_test,ypred)</pre> |
| [76]: | classification_report(y_test,ypred) ' precision recall f1-score support\n\n 0 0.50 1.00 0.67 1\n 1 1.00 0.67 0.83 0.73 4\nweighted avg 0.88 0.75 0.81 4\n' from sklearn.metrics import confusion_matrix,accuracy_score |
| [78]: | <pre>from sklearn.metrics import confusion_matrix, accuracy_score print("Confusion Matrix:", confusion_matrix(y_test, ypred)) Confusion Matrix: [[1 0]</pre> |
| [79]: | <pre>print("Accuracy:", accuracy_score(y_test, ypred)) Accuracy: 0.75</pre> |
| []: | |