```
In [1]: import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         import numpy as np
         from sklearn.model_selection import KFold
from sklearn.model_selection import cross_val_score
         from sklearn.ensemble import BaggingClassifier
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.ensemble import AdaBoostClassifier
         from sklearn.linear_model import LogisticRegression
         from sklearn.svm import SVC
         from sklearn.ensemble import VotingClassifier
         from sklearn import preprocessing
In [2]:
         data=pd.read_csv('Fraud_check.csv')
         data
Out[2]:
              Undergrad Marital.Status Taxable.Income City.Population Work.Experience
                                                                                   Urban
                    NO
                               Single
                                              68833
                                                            50047
                                                                               10
                                                                                     YES
                   YES
                                              33700
                                                            134075
                                                                                18
                                                                                     YES
                             Divorced
           2
                    NO
                              Married
                                              36925
                                                            160205
                                                                                30
                                                                                     YES
                   YES
                               Single
                                              50190
                                                            193264
                                                                                15
                                                                                     YES
           4
                    NO
                              Married
                                              81002
                                                            27533
                                                                                28
                                                                                      NO
         595
                   YES
                             Divorced
                                              76340
                                                            39492
                                                                                 7
                                                                                     YES
         596
                   YES
                             Divorced
                                              69967
                                                             55369
                                                                                 2
                                                                                     YES
         597
                    NO
                             Divorced
                                              47334
                                                            154058
                                                                                 0
                                                                                     YES
         598
                   YES
                              Married
                                              98592
                                                            180083
                                                                                17
                                                                                      NO
         599
                    NΩ
                             Divorced
                                              96519
                                                            158137
                                                                                16
                                                                                      NO
        600 rows × 6 columns
```

In [3]: data.info()

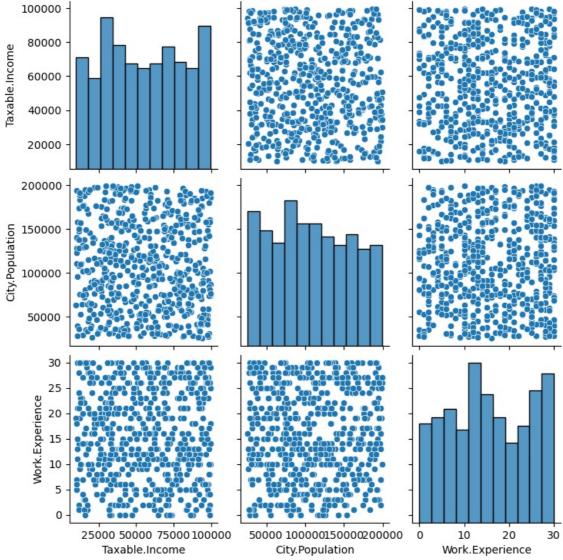
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 600 entries, 0 to 599
Data columns (total 6 columns):

Non-Null Count Dtype # Column 0 600 non-null Undergrad object Marital.Status 600 non-null 1 object Taxable.Income 600 non-null int64 3 City.Population 600 non-null int64 4 Work.Experience 600 non-null int64 Urban 600 non-null object

dtypes: int64(3), object(3)
memory usage: 28.2+ KB

In [4]: sns.pairplot(data)

Out[4]: <seaborn.axisgrid.PairGrid at 0x2a112711490>



```
In [6]: sns.heatmap(data.isnull(),cmap='Reds')
Out[6]: <AxesSubplot:>
In [7]: plt.figure(figsize=(20,10))
sns.heatmap(data.corr(),annot=True)
Out[7]: <AxesSubplot:>
In [8]: label_encoder = preprocessing.LabelEncoder()
data['Undergrad']= label_encoder.fit_transform(data['Undergrad'])
data['Urban']= label_encoder.fit_transform(data['Urban'])
data['Marital.Status']= label_encoder.fit_transform(data['Marital.Status'])
```

```
In [9]: data
 Out[9]:
               Undergrad Marital.Status Taxable.Income City.Population Work.Experience Urban
            0
                      0
                                   2
                                              68833
                                                            50047
                                                                              10
                                                                                      1
            1
                      1
                                   0
                                              33700
                                                           134075
                                                                              18
            2
                      0
                                   1
                                              36925
                                                           160205
                                                                              30
            3
                      1
                                   2
                                              50190
                                                           193264
                                                                              15
                                                                                      1
            4
                      0
                                   1
                                              81002
                                                            27533
                                                                              28
                                                                                      0
                                   0
                                                                               7
          595
                      1
                                              76340
                                                            39492
                                                                                      1
          596
                      1
                                   0
                                              69967
                                                            55369
                                                                               2
                      0
                                   0
                                              47334
                                                                               0
          597
                                                           154058
          598
                                              98592
                                                           180083
                                                                                      0
                                                                              17
          599
                      0
                                   0
                                              96519
                                                           158137
                                                                              16
                                                                                      0
         600 rows × 6 columns
In [10]: data['Status'] = data['Taxable.Income'].apply(lambda Income: 'Risky' if Income <= 30000 else 'Good')</pre>
In [11]: data['Status']= label encoder.fit transform(data['Status'])
In [12]:
          data
               Undergrad Marital.Status Taxable.Income City.Population Work.Experience
                                                                                  Urban Status
            0
                      0
                                   2
                                                                                            0
                                              68833
                                                            50047
                                                                              10
                      1
                                   0
                                              33700
                                                           134075
                                                                              18
                                                                                            0
            2
                      0
                                   1
                                              36925
                                                           160205
                                                                              30
                                                                                             0
                                   2
            3
                      1
                                              50190
                                                           193264
                                                                              15
                                                                                            0
            4
                      0
                                   1
                                              81002
                                                            27533
                                                                              28
                                                                                      0
                                                                                             0
                      1
                                   0
                                                                               7
                                                                                      1
                                                                                            0
          595
                                              76340
                                                            39492
          596
                      1
                                   0
                                              69967
                                                            55369
                                                                               2
                                                                                            0
          597
                      0
                                   0
                                              47334
                                                           154058
                                                                               0
                                                                                      1
                                                                                            0
          598
                                              98592
                                                           180083
                                                                              17
                                                                                      0
                                                                                            0
          599
                      0
                                   0
                                              96519
                                                           158137
                                                                              16
                                                                                      0
                                                                                             0
         600 rows × 7 columns
In [13]: data.Status.unique()
          array([0, 1])
Out[13]:
In [14]:
          x=data.iloc[:,0:4]
          y=data['Status']
In [15]:
          num trees = 100
           seed=8
          kfold = KFold(n splits=100, shuffle = True, random state=seed)
          model = AdaBoostClassifier(n_estimators=num_trees, random_state=seed)
          results = cross_val_score(model, x,y, cv=kfold)
          print(results.mean())
          0.9983333333333334
In [16]:
          kfold = KFold(n_splits=10,shuffle=True, random_state=8)
          estimators = []
          model1 = LogisticRegression(max_iter=100)
          estimators.append(('logistic', model1))
          model2 = DecisionTreeClassifier()
          estimators.append(('cart', model2))
          model3 = SVC()
          estimators.append(('svm', model3))
          ensemble = VotingClassifier(estimators)
          results = cross_val_score(ensemble, x, y, cv=kfold)
          print(results.mean())
          0.983333333333333
In [17]:
          num trees = 100
          max_features = 3
```

```
kfold = KFold(n_splits=10, shuffle= True ,random_state=8)
model = RandomForestClassifier(n_estimators=num_trees, max_features)
results = cross_val_score(model, x, y, cv=kfold)
print(results.mean())

0.9983333333333334

In [18]:
num_trees = 100
seed=8
kfold = KFold(n_splits=100, shuffle = True, random_state=seed)
model = AdaBoostClassifier(n_estimators=num_trees, random_state=seed)
results = cross_val_score(model, x,y, cv=kfold)
print(results.mean())

0.99833333333333334
In []:
```

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