РОССИЙСКИЙ УНИВЕРСИТЕТ ДРУЖБЫ НАРОДОВ

Факультет физико-математических и естественных наук

Кафедра математического моделирования и искусственного интеллекта

ОТЧЕТ ПО ЛАБОРАТОРНОЙ РАБОТЕ № 6

Дисциплина: Интеллектуальный анализ данных

Студент: Генералов Даниил

Группа: НПИбд-01-21

Москва 2024

- 1. Считайте заданный набор данных из репозитария UCI, включая указанный в индивидуальном задании столбец с метками классов..
- 2. Если среди меток класса имеются пропущенные значения, то удалите записи с пропущенными метками класса. Преобразуйте категориальные признаки в числовые при помощи кодирования меток (label encoding). Если в признаках имеются пропущенные значения, то замените пропущенные значения, используя метод, указанный в индивидуальном задании. Если в признаках пропущенных значений нет, то удалите из набора данных записи, идентифицированные как выбросы при помощи метода кластеризации DBSCAN.
- 3. Используя метод снижения размерности данных, указанный в индивидуальном задании, определите и оставьте в наборе данных не более четырех признаков.
- 4. Нормализуйте оставшиеся признаки набора данных методом, указанным в индивидуальном задании.
- 5. Визуализируйте набор данных в виде точек в трехмерном пространстве, отображая точки разных классов разными цветами. При визуализации набора данных используйте три признака с наиболее высокой оценкой важности. В качестве подписей осей используйте названия признаков. В подписи рисунка укажите название набора данных. Создайте легенду набора данных.

- Разбейте набор данных на обучающую и тестовую выборки. Создайте и обучите классификатор на основе деревьев решений с глубиной дерева не более 4, определите долю верных ответов на тестовой выборке и визуализируйте границу принятия решений и построенное дерево решений. При визуализации границы принятия решений используйте два признака с наиболее высокой оценкой важности.
 - 7.
 - 8.
 - 9.

Вариант 10

Automobile Data Set

Название файла: imports-85.data

Ссылка: http://archive.ics.uci.edu/ml/datasets/Automobile

Класс: num-of-doors (столбец No 6)

Метод обработки пропущенных значений - медиана признака

Метод нормализации признаков - нормировка по норме тах

Алгоритм снижения размерности данных - рекурсивное исключение признаков (RFE)

Дополнительные базовые классификаторы:

- наивный байесовский классификатор
- классификатор ближайших соседей (к-во соседей = 3)

Комбинированный классификатор: VotingClassifier

Ансамблевые классификаторы: BaggingClassifier, GradientBoostingClassifier

Показатель качества модели - коэффициент Жаккара (jaccard)

1. открыть базу данных и прочитать значения

```
# fetch dataset
        automobile= fetch ucirepo(id=10)
In [2]: automobile['data'].keys()
Out[2]: dict_keys(['ids', 'features', 'targets', 'original', 'headers'])
In [3]: automobile['data']['features']
Out[3]:
                       highway-
                                        peak-
                                                             compression-
                price
                                               horsepower
                                                                            stroke bore
                                                                     ratio
           0 13495.0
                             27
                                    21 5000.0
                                                      111.0
                                                                       9.0
                                                                              2.68
                                                                                    3.47
           1 16500.0
                             27
                                   21 5000.0
                                                      111.0
                                                                       9.0
                                                                              2.68
                                                                                    3.47
           2 16500.0
                                    19 5000.0
                              26
                                                      154.0
                                                                       9.0
                                                                              3.47
                                                                                    2.68
           3 13950.0
                              30
                                    24 5500.0
                                                      102.0
                                                                              3.40
                                                                                    3.19
                                                                      10.0
           4 17450.0
                              22
                                                                       8.0
                                    18 5500.0
                                                      115.0
                                                                              3.40
                                                                                    3.19
        200 16845.0
                                    23 5400.0
                                                      114.0
                                                                       9.5
                                                                              3.15
                              28
                                                                                    3.78
        201 19045.0
                              25
                                    19 5300.0
                                                      160.0
                                                                       8.7
                                                                              3.15
                                                                                    3.78
        202 21485.0
                              23
                                    18 5500.0
                                                      134.0
                                                                       8.8
                                                                              2.87
                                                                                    3.58
        203 22470.0
                                   26 4800.0
                                                                                    3.01
                              27
                                                      106.0
                                                                      23.0
                                                                              3.40
        204 22625.0
                              25
                                    19 5400.0
                                                      114.0
                                                                       9.5
                                                                              3.15
                                                                                    3.78
        205 rows \times 25 columns
In [4]: | automobile['data']['features']['num-of-doors'].unique()
Out[4]: array([ 2., 4., nan])
In [5]: data = automobile['data']['features']
```

2. предобработка данных

```
In [6]: KLASS = 'num-of-doors'

data = data.dropna(subset=[KLASS])
    data
```

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v	u			v	- 1	

		price	highway- mpg	city- mpg	peak- rpm	horsepower	compression- ratio	stroke	bore
	0	13495.0	27	21	5000.0	111.0	9.0	2.68	3.47
	1	16500.0	27	21	5000.0	111.0	9.0	2.68	3.47
	2	16500.0	26	19	5000.0	154.0	9.0	3.47	2.68
	3	13950.0	30	24	5500.0	102.0	10.0	3.40	3.19
	4	17450.0	22	18	5500.0	115.0	8.0	3.40	3.19
									•••
2	200	16845.0	28	23	5400.0	114.0	9.5	3.15	3.78
2	201	19045.0	25	19	5300.0	160.0	8.7	3.15	3.78
2	202	21485.0	23	18	5500.0	134.0	8.8	2.87	3.58
2	203	22470.0	27	26	4800.0	106.0	23.0	3.40	3.01
2	204	22625.0	25	19	5400.0	114.0	9.5	3.15	3.78

203 rows \times 25 columns

```
In [7]: from sklearn.preprocessing import LabelEncoder
KATEGORICAL = ['fuel-system', 'engine-location', 'drive-wheels', 'engine-type ENCODERS = []
for k in KATEGORICAL:
    encoder = LabelEncoder()
    data[k] = encoder.fit_transform(data[k])
    ENCODERS.append(encoder)
```

```
/tmp/ipykernel 53338/2462000071.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/
stable/user guide/indexing.html#returning-a-view-versus-a-copy
  data[k] = encoder.fit transform(data[k])
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Out[7]:

	price	highway- mpg	city- mpg	peak- rpm	horsepower	compression- ratio	stroke	bore
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204	22625.0	25	19	5400.0	114.0	9.5	3.15	3.78

203 rows \times 25 columns

```
In [8]: for col in data.columns:
    data[col] = data[col].fillna(data[col].median())

data
```

```
/tmp/ipykernel 53338/3630728420.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
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Out[8]:

:		price	highway- mpg	city- mpg	peak- rpm	horsepower	compression- ratio	stroke	bore
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2	203	22470.0	27	26	4800.0	106.0	23.0	3.40	3.01
2	204	22625.0	25	19	5400.0	114.0	9.5	3.15	3.78

203 rows \times 25 columns

3. снижение размерности

```
In [9]: from sklearn.feature_selection import RFE
from sklearn.linear_model import SGDClassifier

est = SGDClassifier()
    rfe = RFE(estimator=est, n_features_to_select=4)

X = data.drop(KLASS, axis=1)

rfe.fit(X, data[KLASS])

print(rfe.support_)
```

[False False False

```
In [10]: Xreduced = X[X.columns[rfe.support_]]
    Xreduced
```

Out[10]:		horsepower	engine-size	curb-weight	normalized-losses
	0	111.0	130	2548	115.0
	1	111.0	130	2548	115.0
	2	154.0	152	2823	115.0
	3	102.0	109	2337	164.0
	4	115.0	136	2824	164.0
	200	114.0	141	2952	95.0
	201	160.0	141	3049	95.0
	202	134.0	173	3012	95.0
	203	106.0	145	3217	95.0
	204	114.0	141	3062	95.0

203 rows × 4 columns

4. нормализация данных

Out[11]:		horsepower	engine-size	curb-weight	normalized-losses
	0	-0.614583	-0.601227	-0.373340	-0.550781
	1	-0.614583	-0.601227	-0.373340	-0.550781
	2	-0.465278	-0.533742	-0.305706	-0.550781
	3	-0.645833	-0.665644	-0.425234	-0.359375
	4	-0.600694	-0.582822	-0.305460	-0.359375
	200	-0.604167	-0.567485	-0.273979	-0.628906
	201	-0.44444	-0.567485	-0.250123	-0.628906
	202	-0.534722	-0.469325	-0.259223	-0.628906
	203	-0.631944	-0.555215	-0.208805	-0.628906
	204	-0.604167	-0.567485	-0.246926	-0.628906

203 rows × 4 columns

5. визуализация

```
In [12]: rfe_more = RFE(estimator=SGDClassifier(), n_features_to_select=3)
    rfe_more.fit(Xreduced, data[KLASS])

    rfe_most = RFE(estimator=SGDClassifier(), n_features_to_select=2)
    rfe_most.fit(Xreduced, data[KLASS])

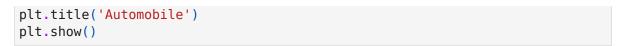
    Xreduced_more = Xreduced[Xreduced.columns[rfe_more.support_]]
    Xreduced_most = Xreduced[Xreduced.columns[rfe_most.support_]]

    Xreduced_more
```

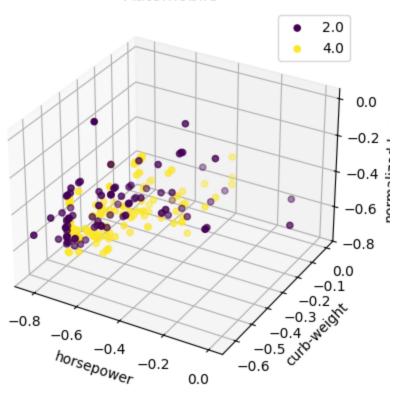
Out[12]: horser

	horsepower	curb-weight	normalized-losses
0	-0.614583	-0.373340	-0.550781
1	-0.614583	-0.373340	-0.550781
2	-0.465278	-0.305706	-0.550781
3	-0.645833	-0.425234	-0.359375
4	-0.600694	-0.305460	-0.359375
200	-0.604167	-0.273979	-0.628906
201	-0.44444	-0.250123	-0.628906
202	-0.534722	-0.259223	-0.628906
203	-0.631944	-0.208805	-0.628906
204	-0.604167	-0.246926	-0.628906

203 rows \times 3 columns







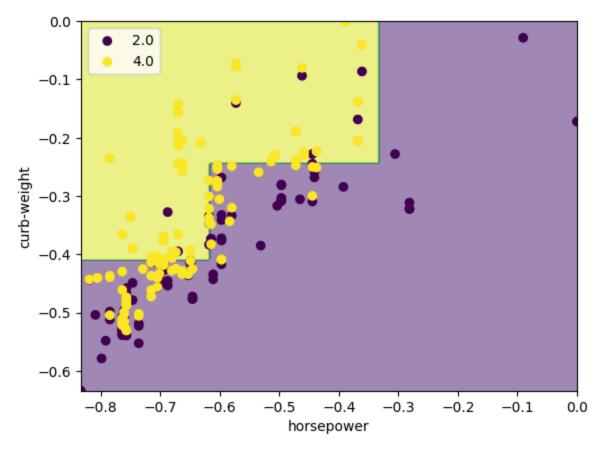
6. классификация

```
In [16]: train_ok = (clf.predict(X_train) == y_train).sum()
   test_ok = (clf.predict(X_test) == y_test).sum()
   print("Train OK:", train_ok, "/", len(y_train), "=", train_ok/len(y_train))
   print("Test OK:", test_ok, "/", len(y_test), "=", test_ok/len(y_test))
```

Train OK: 139 / 162 = 0.8580246913580247Test OK: 26 / 41 = 0.6341463414634146

```
In [17]: import pandas as pd
         import numpy as np
         from sklearn.inspection import DecisionBoundaryDisplay
         fig = plt.figure()
         ax = fig.add subplot()
         cols = Xreduced most.columns
         xx0, xx1 = np.meshgrid(np.linspace(min(Xreduced most[cols[0]]), max(Xreduced))
         least important col = list(set(Xreduced more.columns) - set(cols))[0]
         least important col avg = Xreduced more[least important col].mean()
         Xmeshgrid = pd.DataFrame(np.array([xx0.ravel(), xx1.ravel(), [least importar
         response = clf.predict(Xmeshgrid)
         response = response.reshape(xx0.shape)
         DecisionBoundaryDisplay(xx0=xx0, xx1=xx1, response=response, xlabel=cols[0],
         for k in set(data[KLASS]):
             Xreduced most subset = Xreduced most[data[KLASS] == k]
             ax.scatter(Xreduced most subset[cols[0]], Xreduced most subset[cols[1]],
         ax.legend()
```

Out[17]: <matplotlib.legend.Legend at 0x7f09d5c70750>



7. комбинирование классификаторов

```
In [18]: from sklearn.naive bayes import GaussianNB
         clf2 = GaussianNB()
         clf2.fit(X train, y train)
         train ok = (clf2.predict(X train) == y train).sum()
         test ok = (clf2.predict(X test) == y test).sum()
         print("Train OK:", train_ok, "/", len(y_train), "=", train_ok/len(y_train))
         print("Test OK:", test ok, "/", len(y test), "=", test ok/len(y test))
        Train 0K: 107 / 162 = 0.6604938271604939
        Test 0K: 21 / 41 = 0.5121951219512195
In [19]: \# k nearest neighbors (k=3)
         from sklearn.neighbors import KNeighborsClassifier
         clf3 = KNeighborsClassifier(n_neighbors=3)
         clf3.fit(X train, y train)
         train ok = (clf3.predict(X train) == y train).sum()
         test ok = (clf3.predict(X test) == y test).sum()
         print("Train OK:", train_ok, "/", len(y_train), "=", train_ok/len(y_train))
         print("Test OK:", test ok, "/", len(y test), "=", test ok/len(y test))
        Train 0K: 146 / 162 = 0.9012345679012346
        Test 0K: 31 / 41 = 0.7560975609756098
In [20]: from sklearn.ensemble import VotingClassifier
         clf4 = VotingClassifier([
             ('decision-tree', clf),
             ('naive-bayes', clf2),
             ('knn', clf3)
         ])
         clf4.fit(X train, y train)
         train ok = (clf4.predict(X train) == y train).sum()
         test ok = (clf4.predict(X test) == y test).sum()
         print("Train OK:", train_ok, "/", len(y_train), "=", train_ok/len(y train))
         print("Test OK:", test_ok, "/", len(y_test), "=", test_ok/len(y_test))
        Train 0K: 141 / 162 = 0.8703703703703703
        Test 0K: 26 / 41 = 0.6341463414634146
In [21]: from sklearn.metrics import jaccard score
         print("jaccard score:")
         print("Decision tree:", jaccard_score(y_test, clf.predict(X_test), average="
         print("Naive Bayes :", jaccard score(y test, clf2.predict(X test), average=
```

```
print("KNN : ", jaccard_score(y_test, clf3.predict(X_test), average=
print("Voting : ", jaccard_score(y_test, clf4.predict(X_test), average=

jaccard_score:
Decision tree: 0.46593776282590416
Naive Bayes : 0.3448110951697752
KNN : 0.6086303939962476
Voting : 0.4662288930581613
```

8. ансамблевый классификатор

```
In [22]: from sklearn.ensemble import BaggingClassifier
        clf5 = BaggingClassifier(n estimators=1000)
        clf5.fit(X train, y train)
        train ok = (clf5.predict(X_train) == y_train).sum()
        test ok = (clf5.predict(X test) == y test).sum()
        print("Train OK:", train_ok, "/", len(y_train), "=", train_ok/len(y_train))
        print("Test OK:", test_ok, "/", len(y_test), "=", test_ok/len(y_test))
       Train 0K: 162 / 162 = 1.0
       Test 0K: 33 / 41 = 0.8048780487804879
In [23]: # gradient boosting
        from sklearn.ensemble import GradientBoostingClassifier
        clf6 = GradientBoostingClassifier(n estimators=1000)
        clf6.fit(X train, y train)
        train ok = (clf6.predict(X train) == y train).sum()
        test ok = (clf6.predict(X test) == y test).sum()
        print("Train OK:", train_ok, "/", len(y_train), "=", train_ok/len(y_train))
        print("Test OK:", test_ok, "/", len(y_test), "=", test_ok/len(y_test))
       Train 0K: 162 / 162 = 1.0
       Test 0K: 29 / 41 = 0.7073170731707317
In [24]: print("jaccard score:")
        print("Decision tree:", jaccard_score(y_test, clf.predict(X_test), average='
        print("Naive Bayes :", jaccard score(y test, clf2.predict(X test), average=
        jaccard score:
       Decision tree: 0.46593776282590416
       Naive Bayes : 0.3448110951697752
             : 0.6086303939962476
       KNN
       Voting
                  : 0.4662288930581613
       Bagging : 0.6741463414634147
       Gradient : 0.5488501742160279
```

9. границы ансамблевых классификаторов

```
In [25]: fig = plt.figure()
    ax = fig.add_subplot()

cols = Xreduced_most.columns

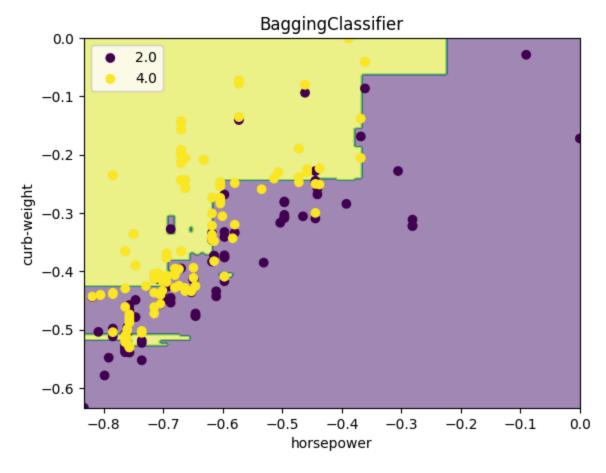
xx0, xx1 = np.meshgrid(np.linspace(min(Xreduced_most[cols[0]]), max(Xreduced_least_important_col = list(set(Xreduced_more.columns) - set(cols))[0]
    least_important_col_avg = Xreduced_more[least_important_col].mean()

Xmeshgrid = pd.DataFrame(np.array([xx0.ravel(), xx1.ravel(), [least_important_response = clf5.predict(Xmeshgrid)
    response = response.reshape(xx0.shape)

DecisionBoundaryDisplay(xx0=xx0, xx1=xx1, response=response, xlabel=cols[0],

for k in set(data[KLASS]):
    Xreduced_most_subset = Xreduced_most[data[KLASS] == k]
    ax.scatter(Xreduced_most_subset[cols[0]], Xreduced_most_subset[cols[1]],
ax.legend()
    plt.title("BaggingClassifier")
```

Out[25]: Text(0.5, 1.0, 'BaggingClassifier')



```
In [26]:
         fig = plt.figure()
         ax = fig.add subplot()
         cols = Xreduced most.columns
         xx0, xx1 = np.meshgrid(np.linspace(min(Xreduced most[cols[0]]), max(Xreduced
         least important col = list(set(Xreduced more.columns) - set(cols))[0]
         least important col avg = Xreduced more[least important col].mean()
         Xmeshgrid = pd.DataFrame(np.array([xx0.ravel(), xx1.ravel(), [least importar
         response = clf6.predict(Xmeshgrid)
         response = response.reshape(xx0.shape)
         DecisionBoundaryDisplay(xx0=xx0, xx1=xx1, response=response, xlabel=cols[0],
         for k in set(data[KLASS]):
             Xreduced most subset = Xreduced most[data[KLASS] == k]
             ax.scatter(Xreduced most subset[cols[0]], Xreduced most subset[cols[1]],
         ax.legend()
         plt.title("GradientBoostingClassifier")
```

Out[26]: Text(0.5, 1.0, 'GradientBoostingClassifier')

