### Московский государственный технический университет им. Н.Э. Баумана Кафедра «Системы обработки информации и управления»

# Лабораторная работа №4 по дисциплине «Методы машинного обучения» на тему

«Подготовка обучающей и тестовой выборки, кросс-валидация и подбор гиперпараметров на примере метода ближайших соседей.»

Выполнила: студентка группы ИУ5-24М Горбовцова К.М.

```
In [2]: | import numpy as np
                        import pandas as pd
In [3]: from typing import Dict, Tuple
In [4]: from scipy import stats
In [5]: | from sklearn.datasets import load_iris, load boston
                        from sklearn.model selection import train test split
                        from sklearn.neighbors import KNeighborsRegressor, KNeighborsClassifie
                        from sklearn.metrics import accuracy score
                        from sklearn.metrics import balanced accuracy score
                        from sklearn.metrics import plot_confusion_matrix
                        from sklearn.metrics import precision_score, recall_score, f1_score, c
                        from sklearn.metrics import confusion matrix
                        from sklearn.model selection import GridSearchCV
                        from sklearn.model selection import cross val score, cross validate
                        from sklearn.model selection import KFold, RepeatedKFold, LeaveOneOut,
                        from sklearn.model selection import learning curve, validation curve
                        from sklearn.metrics import mean absolute error, mean squared erro
                        from sklearn.metrics import roc curve, roc auc score
In [6]:
                       import seaborn as sns
                        import matplotlib.pyplot as plt
In [7]:
In [8]: %matplotlib inline
                        sns.set(style="ticks")
In [9]: data = pd.read csv("heart.csv")
```

# In [10]: data.info(5)

<class 'pandas.core.frame.DataFrame'> RangeIndex: 303 entries, 0 to 302 Data columns (total 14 columns): 303 non-null int64 age sex 303 non-null int64 303 non-null int64 ср 303 non-null int64 trestbps 303 non-null int64 chol fbs 303 non-null int64 restecg 303 non-null int64 303 non-null int64 thalach 303 non-null int64 exang oldpeak 303 non-null float64 slope 303 non-null int64 303 non-null int64 ca thal 303 non-null int64 target 303 non-null int64 dtypes: float64(1), int64(13)

memory usage: 33.2 KB

# In [11]: data.describe()

### Out[11]:

	age	sex	ср	trestbps	chol	fbs	restecg	
count	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	-;
mean	54.366337	0.683168	0.966997	131.623762	246.264026	0.148515	0.528053	
std	9.082101	0.466011	1.032052	17.538143	51.830751	0.356198	0.525860	
min	29.000000	0.000000	0.000000	94.000000	126.000000	0.000000	0.000000	
25%	47.500000	0.000000	0.000000	120.000000	211.000000	0.000000	0.000000	
50%	55.000000	1.000000	1.000000	130.000000	240.000000	0.000000	1.000000	
75%	61.000000	1.000000	2.000000	140.000000	274.500000	0.000000	1.000000	
max	77.000000	1.000000	3.000000	200.000000	564.000000	1.000000	2.000000	1

In [12]: data.corr()

Out[12]:

	age	sex	ср	trestbps	chol	fbs	restecg	thalac
age	1.000000	-0.098447	-0.068653	0.279351	0.213678	0.121308	-0.116211	-0.39852
sex	-0.098447	1.000000	-0.049353	-0.056769	-0.197912	0.045032	-0.058196	-0.04402
ср	-0.068653	-0.049353	1.000000	0.047608	-0.076904	0.094444	0.044421	0.29576
trestbps	0.279351	-0.056769	0.047608	1.000000	0.123174	0.177531	-0.114103	-0.04669
chol	0.213678	-0.197912	-0.076904	0.123174	1.000000	0.013294	-0.151040	-0.00994
fbs	0.121308	0.045032	0.094444	0.177531	0.013294	1.000000	-0.084189	-0.00856
restecg	-0.116211	-0.058196	0.044421	-0.114103	-0.151040	-0.084189	1.000000	0.04412
thalach	-0.398522	-0.044020	0.295762	-0.046698	-0.009940	-0.008567	0.044123	1.00000
exang	0.096801	0.141664	-0.394280	0.067616	0.067023	0.025665	-0.070733	-0.37881
oldpeak	0.210013	0.096093	-0.149230	0.193216	0.053952	0.005747	-0.058770	-0.34418
slope	-0.168814	-0.030711	0.119717	-0.121475	-0.004038	-0.059894	0.093045	0.38678
са	0.276326	0.118261	-0.181053	0.101389	0.070511	0.137979	-0.072042	-0.21317
thal	0.068001	0.210041	-0.161736	0.062210	0.098803	-0.032019	-0.011981	-0.09643
target	-0.225439	-0.280937	0.433798	-0.144931	-0.085239	-0.028046	0.137230	0.42174

```
In [13]: np.unique(data.target)
```

Out[13]: array([0, 1])

```
In [14]: target = data.iloc[:, -1]
```

```
In [15]: data_data = data.iloc[:, 0:-1]
```

- In [16]: target.shape
- Out[16]: (303,)
- In [17]: data\_data.shape
- Out[17]: (303, 13)

```
In [19]: heart X train.shape, heart y train.shape
Out[19]: ((151, 13), (151,))
In [20]:
        heart X test.shape, heart y test.shape
Out[20]: ((152, 13), (152,))
In [21]: np.unique(heart y train), np.unique(heart y test)
Out[21]: (array([0, 1]), array([0, 1]))
In [22]: cl1 1 = KNeighborsClassifier(n neighbors=2)
         cll_1.fit(heart_X_train, heart_y_train)
         target1 1 = cl1 1.predict(heart X test)
         len(target1_1), target1_1
Out[22]: (152, array([0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1
         , 0, 0, 0,
                 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0,
         0, 0,
                 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1, 1,
         1, 1,
                 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0,
         1, 0,
                 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0,
         1, 0,
                 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1,
         0, 0,
                 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1]
         )
```

```
In [23]:
         cl1 2 = KNeighborsClassifier(n neighbors=10)
         cl1 2.fit(heart X train, heart y train)
         target1 2 = cl1 2.predict(heart X test)
         len(target1_2), target1_2
Out[23]: (152, array([1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1
         , 0, 1, 1,
                  0, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0,
         0,0,
                 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1,
         1, 0,
                 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0,
         0, 0,
                 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 1, 1, 1, 0, 1, 0,
         1, 0,
                 1, 0, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1,
         1, 0,
                 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1]
         )
         accuracy_score(heart_y_test, target1 1)
In [24]:
Out[24]: 0.6052631578947368
In [25]: accuracy score(heart y test, target1 2)
Out[25]: 0.625
In [26]:
         cl1_3 = KNeighborsClassifier(n_neighbors=30)
         cl1 3.fit(heart X train, heart y train)
         target1_3 = cl1_3.predict(heart_X_test)
         len(target1 3), target1 3
Out[26]: (152, array([1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1
         , 0, 0, 0,
                  1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1,
         1, 0,
                 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1,
         1, 1,
                 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0,
         0, 0,
                 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0,
         1, 1,
                 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1,
         1, 1,
                 0, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1]
         )
```

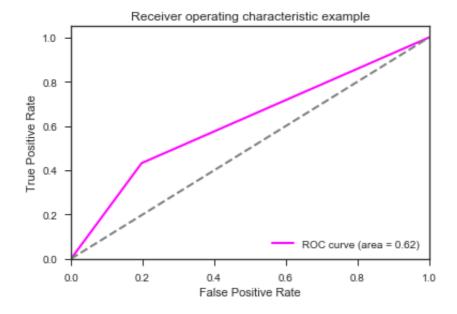
In [27]: accuracy score(heart y test, target1 3)

Out[28]: array([0.67391304, 0.67391304, 0.62214834, 0.58823529, 0.63636364])

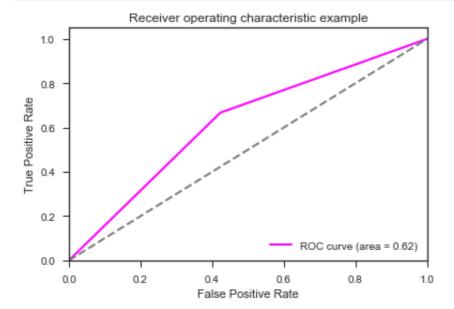
```
In [29]:
        scores = cross val score(KNeighborsClassifier(n neighbors=50),
                              data data, target,
                              cv=LeaveOneOut())
        scores, np.mean(scores)
1., 1.,
               0., 1., 1., 1., 1., 1., 0., 1., 1., 0., 0., 1., 1., 0.,
        1., 0.,
               0., 1., 1., 1., 0., 0., 0., 1., 1., 0., 1., 1., 1., 1., 0.,
        1., 1.,
               0., 1., 1., 1., 1., 1., 1., 1., 1., 0., 1., 1., 0., 1., 1.,
        1., 1.,
               1., 0.,
               1., 0., 1., 1., 0., 1., 1., 1., 1., 1., 0., 1., 1., 0., 1.,
        1., 0.,
               1., 1., 1., 0., 0., 1., 1., 1., 0., 1., 0., 1., 1., 1., 1.,
        1., 1.,
               1., 0., 1., 1., 1., 1., 1., 1., 1., 0., 1., 1., 1., 1.,
        1., 1.,
               0., 1., 1., 0., 1., 1., 1., 1., 0., 0., 0., 1., 1., 1., 0.,
        0., 1.,
               0., 0.,
               1., 0., 0., 0., 1., 1., 0., 0., 1., 1., 1., 1., 0., 0., 1.,
        0., 0.,
               1., 0., 0., 1., 1., 1., 1., 0., 1., 0., 0., 1., 0., 0., 1.,
        1., 1.,
               1., 0., 1., 0., 0., 0., 0., 1., 0., 1., 1., 1., 1., 1., 1.,
        0., 0.,
               1., 0., 1., 1., 1., 1., 0., 0., 1., 0., 1., 1., 1., 1., 0.,
        0., 0.,
               0., 0., 1., 1., 1., 1., 0., 0., 1., 0., 0., 1., 1., 1.,
        1., 1.,
               1., 1., 1., 0., 0., 0., 0., 1., 0., 1., 0., 1., 0., 1., 1.,
        1., 1.,
               1., 0., 1., 0., 1., 1., 0., 1., 1., 0., 0., 0., 0., 1., 0.,
        0., 1.,
               1., 0., 1., 0., 0., 0., 1., 0., 1., 1., 1., 0., 0., 0.]),
         0.6633663366336634)
In [30]: | fpr, tpr, thresholds = roc_curve(heart_y_test, target1_2,
                                     pos label=1)
        fpr, tpr, thresholds
                        , 0.42253521, 1.
Out[30]: (array([0.
                                              1),
                        , 0.66666667, 1.
         array([0.
                                              1),
         array([2, 1, 0]))
```

```
In [62]:
         def draw roc curve(y true, y score, pos label, average):
             fpr, tpr, thresholds = roc curve(y true, y score,
                                               pos_label=pos_label)
             roc_auc_value = roc_auc_score(y_true, y_score, average=average)
             plt.figure()
             lw = 2
             plt.plot(fpr, tpr, color='magenta',
                      lw=lw, label='ROC curve (area = %0.2f)' % roc auc value)
             plt.plot([0, 1], [0, 1], color='gray', lw=lw, linestyle='--')
             plt.xlim([0.0, 1.0])
             plt.ylim([0.0, 1.05])
             plt.xlabel('False Positive Rate')
             plt.ylabel('True Positive Rate')
             plt.title('Receiver operating characteristic example')
             plt.legend(loc="lower right")
             plt.show()
```

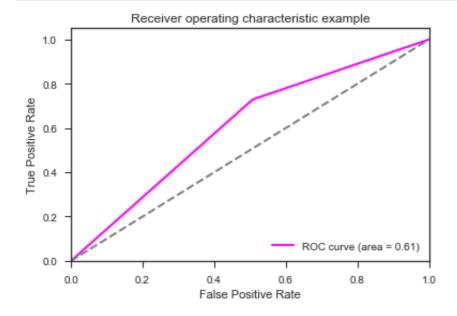
# In [63]: draw\_roc\_curve(heart\_y\_test, target1\_1, pos\_label=1, average='micro')



In [64]: draw\_roc\_curve(heart\_y\_test, target1\_2, pos\_label=1, average='micro')



In [65]: draw\_roc\_curve(heart\_y\_test, target1\_3, pos\_label=1, average='micro')



In [67]: scores

Out[67]: array([0.62376238, 0.6039604 , 0.66336634])

In [68]: np.mean(scores)

Out[68]: 0.6303630363036303

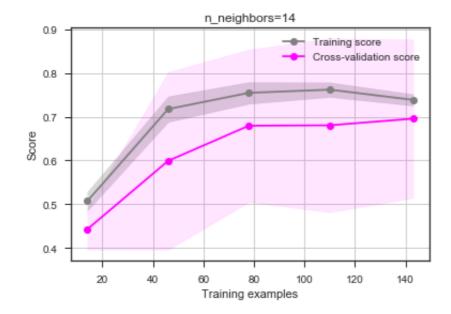
```
In [69]: n range = np.array(range(5,55,5))
         tuned parameters = [{'n neighbors': n range}]
         tuned parameters
Out[69]: [{'n neighbors': array([ 5, 10, 15, 20, 25, 30, 35, 40, 45, 50])}]
In [70]:
         %%time
         clf_gs = GridSearchCV(KNeighborsClassifier(), tuned_parameters, cv=5,
         clf gs.fit(heart X train, heart y train)
         CPU times: user 223 ms, sys: 4.32 ms, total: 227 ms
         Wall time: 233 ms
In [71]: clf gs.best estimator
Out[71]: KNeighborsClassifier(algorithm='auto', leaf size=30, metric='minkows
         ki',
                              metric params=None, n jobs=None, n neighbors=35
         p=2
                              weights='uniform')
In [72]: clf gs.best score
Out[72]: 0.6823655913978494
In [73]: clf gs.best params
Out[73]: {'n_neighbors': 35}
In [74]: plt.plot(n range, clf gs.cv results ['mean test score'])
Out[74]: [<matplotlib.lines.Line2D at 0x11cb35128>]
          0.680
          0.675
          0.670
          0.665
          0.660
```

10

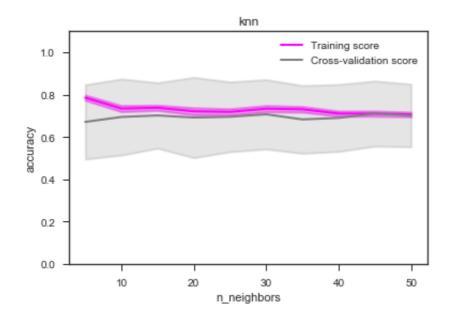
20

30

```
In [75]: def plot learning curve(estimator, title, X, y, ylim=None, cv=None,
                                 n jobs=None, train sizes=np.linspace(.1, 1.0,
             plt.figure()
             plt.title(title)
             if ylim is not None:
                 plt.ylim(*ylim)
             plt.xlabel("Training examples")
             plt.ylabel("Score")
             train_sizes, train_scores, test_scores = learning_curve(
                 estimator, X, y, cv=cv, n jobs=n jobs, train sizes=train sizes
             train scores mean = np.mean(train scores, axis=1)
             train scores std = np.std(train scores, axis=1)
             test scores mean = np.mean(test scores, axis=1)
             test scores std = np.std(test scores, axis=1)
             plt.grid()
             plt.fill_between(train_sizes, train_scores_mean - train_scores_std
                              train scores mean + train scores std, alpha=0.3,
                              color="grey")
             plt.fill between(train sizes, test scores mean - test scores std,
                              test_scores_mean + test_scores_std, alpha=0.1, co
             plt.plot(train sizes, train scores mean, 'o-', color="grey",
                      label="Training score")
             plt.plot(train sizes, test scores mean, 'o-', color="magenta",
                      label="Cross-validation score")
             plt.legend(loc="best")
             return plt
```



```
In [79]: def plot validation curve(estimator, title, X, y,
                                   param name, param range, cv,
                                    scoring="accuracy"):
             train scores, test scores = validation curve(
                 estimator, X, y, param name=param name, param range=param range
                 cv=cv, scoring=scoring, n jobs=1)
             train scores mean = np.mean(train scores, axis=1)
             train_scores_std = np.std(train_scores, axis=1)
             test scores mean = np.mean(test scores, axis=1)
             test_scores_std = np.std(test_scores, axis=1)
             plt.title(title)
             plt.xlabel(param name)
             plt.ylabel(str(scoring))
             plt.ylim(0.0, 1.1)
             lw = 2
             plt.plot(param range, train_scores_mean, label="Training score",
                          color="magenta", lw=lw)
             plt.fill between(param range, train scores mean - train scores std
                              train_scores_mean + train_scores_std, alpha=0.4,
                              color="magenta", lw=lw)
             plt.plot(param_range, test_scores_mean, label="Cross-validation sc
                          color="grey", lw=lw)
             plt.fill_between(param_range, test_scores_mean - test_scores_std,
                              test scores mean + test scores std, alpha=0.2,
                              color="grey", lw=lw)
             plt.legend(loc="best")
             return plt
```



```
In [81]: n_range = np.array(range(5,55,5))
  tuned_parameters = [{'n_neighbors': n_range}]
  tuned_parameters
```

Out[81]: [{'n\_neighbors': array([ 5, 10, 15, 20, 25, 30, 35, 40, 45, 50])}]

In [82]: %%time
 clf\_gs = GridSearchCV(KNeighborsClassifier(), tuned\_parameters, cv=Lear
 clf\_gs.fit(data\_data, target)

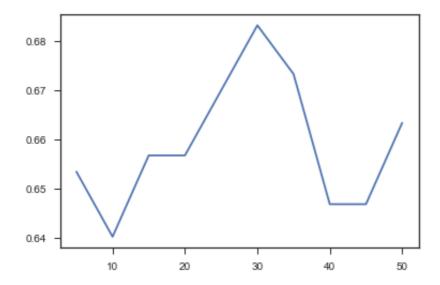
CPU times: user 12.4 s, sys: 52.3 ms, total: 12.4 s Wall time: 13 s

In [83]: clf\_gs.best\_params\_

Out[83]: {'n neighbors': 30}

```
In [84]: plt.plot(n_range, clf_gs.cv_results_['mean_test_score'])
```

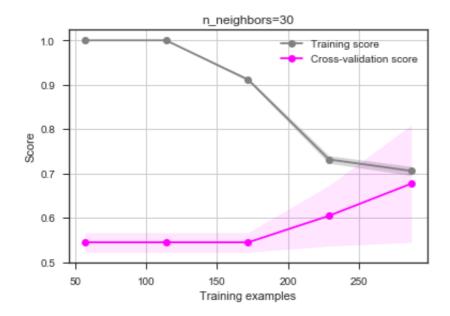
Out[84]: [<matplotlib.lines.Line2D at 0x11c9c2ba8>]



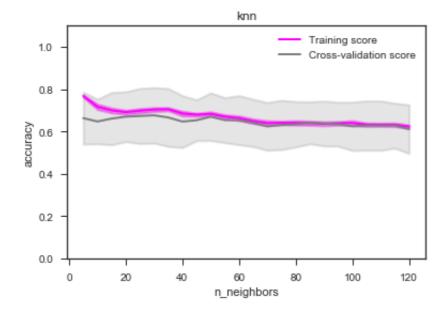
```
In [85]: clf_gs.best_estimator_.fit(heart_X_train, heart_y_train)
    target2_0 = clf_gs.best_estimator_.predict(heart_X_train)
    target2_1 = clf_gs.best_estimator_.predict(heart_X_test)
```

In [86]: accuracy\_score(heart\_y\_train, target2\_0), accuracy\_score(heart\_y\_test,

Out[86]: (0.7284768211920529, 0.618421052631579)



```
In [88]: n_range2 = np.array(range(5,125,5))
```



In [ ]: