Лабораторная работа №3

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

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Задание:

- 1.Выберите набор данных (датасет) для решения задачи классификации или регрессии.
- 2.С использованием метода train_test_split разделите выборку на обучающую и тестовую.
- 3.Обучите модель ближайших соседей для произвольно заданного гиперпараметра К. Оцените качество модели с помощью подходящих для задачи метрик.
- 4.Произведите подбор гиперпараметра К с использованием GridSearchCV и/или RandomizedSearchCV и кросс-валидации, оцените качество оптимальной модели. Желательно использование нескольких стратегий кросс-валидации.
- 5. Сравните метрики качества исходной и оптимальной моделей.

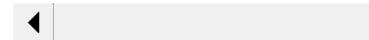
```
In [1]:
from IPython.display import Image
import numpy as np
import pandas as pd
from sklearn.datasets import *
from typing import Dict, Tuple
from sklearn.model selection import train test split
import seaborn as sns
import matplotlib.pyplot as plt
from operator import itemgetter
import matplotlib.ticker as ticker
import math
from sklearn.metrics import accuracy score, balanced accuracy score
from sklearn.metrics import plot confusion matrix
from sklearn.metrics import precision score, recall score, fl score, classification report
from sklearn.metrics import confusion matrix
from sklearn.metrics import mean absolute error, mean squared error, mean squared log error, median absolute e
from sklearn.metrics import roc curve, roc auc score
from sklearn.neighbors import KNeighborsRegressor, KNeighborsClassifier
from sklearn.model_selection import cross_val_score, cross_validate
from sklearn.model selection import KFold, RepeatedKFold, LeaveOneOut, LeavePOut, ShuffleSplit, StratifiedKFold
from sklearn.preprocessing import MinMaxScaler, StandardScaler, Normalizer
from sklearn.model selection import GridSearchCV, RandomizedSearchCV
from sklearn.model selection import learning curve, validation curve
%matplotlib inline
sns.set(style="whitegrid")
```

```
Выборка датасета и ее разделение на тестовую и обучающую
                                                                                                              In [2]:
breast cancer = load breast cancer()
                                                                                                              In [3]:
for x in breast cancer:
    print(x)
data
target
frame
target names
DESCR
feature names
filename
                                                                                                              In [4]:
# Сформируем DataFrame
breast cancer df = pd.DataFrame(data= np.c [breast cancer['data'], breast cancer['target']],
                      columns= list(breast cancer['feature names']) + ['target'])
                                                                                                              In [5]:
breast cancer df
```

Out[5]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	 worst texture	worst perimeter	worst area	sm
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.30010	0.14710	0.2419	0.07871	 17.33	184.60	2019.0	
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.08690	0.07017	0.1812	0.05667	 23.41	158.80	1956.0	
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.19740	0.12790	0.2069	0.05999	 25.53	152.50	1709.0	
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.24140	0.10520	0.2597	0.09744	 26.50	98.87	567.7	
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.19800	0.10430	0.1809	0.05883	 16.67	152.20	1575.0	
•••											 			
564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.1726	0.05623	 26.40	166.10	2027.0	
565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.1752	0.05533	 38.25	155.00	1731.0	
566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.1590	0.05648	 34.12	126.70	1124.0	
567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.2397	0.07016	 39.42	184.60	1821.0	
568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.1587	0.05884	 30.37	59.16	268.6	

569 rows × 31 columns



```
In [6]:
sc = MinMaxScaler()
breast_cancer_sc = sc.fit_transform(breast_cancer.data)
                                                                                                             In [7]:
X_train, X_test, Y_train, Y_test = train_test_split(
breast_cancer_sc, breast_cancer.target, test_size=0.33, random_state=1)
                                                                                                             In [8]:
# Размер обучающей выборки
X train.shape, Y train.shape
                                                                                                            Out[8]:
((381, 30), (381,))
                                                                                                             In [9]:
# Размер тестовой выборки
X test.shape, Y test.shape
                                                                                                            Out[9]:
((188, 30), (188,))
                                                                                                            In [10]:
X_train
                                                                                                           Out[10]:
array([[0.55274741, 0.25059182, 0.53631401, ..., 0.57525773, 0.26197516,
        0.19362456],
       [0.2607317, 0.24146094, 0.24462719, ..., 0.0956701, 0.06938695,
       0.04394595],
       [0.282976 , 0.29015894, 0.27910994, ..., 0.22707904, 0.32367435,
       0.11432507],
       [0.48364807, 0.50084545, 0.48655933, ..., 0.65257732, 0.34456929,
       [0.3336173, 0.3902604, 0.31787713, ..., 0.27364261, 0.13029765,
       0.13859373],
       [0.28628899, 0.29455529, 0.26826066, ..., 0.17226804, 0.08318549,
        0.043618 ]])
                                                                                                            In [11]:
```

Y train

```
array([0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0,
      1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1,
       1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0,
      1, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0,
      0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0,
      0, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 0, 1,
      1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0,
      1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 1, 0,
      1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0,
      1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1,
       1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 0,
       0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1,
      1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0, 0,
      0, 0, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 1, 0,
      0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0,
      0, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0,
       1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1,
       1, 1, 0, 1, 0, 1, 1])
                                                                                                          In [12]:
X_{test}
                                                                                                         Out[12]:
array([[0.36485399, 0.14440311, 0.37613157, ..., 0.38075601, 0.24876799,
       0.24294897],
       [0.29291495, 0.30267163, 0.29154861, ..., 0.71752577, 0.46027991,
       0.41230487],
       [0.28250272, 0.21339195, 0.27192316, ..., 0.3628866, 0.35777646,
       0.26761118],
       [0.60717497, 0.42069665, 0.59574321, ..., 0.82061856, 0.23713779,
       [0.49642671, 0.50625634, 0.49968903, ..., 0.59140893, 0.1172876,
       0.24898334],
       [0.29906763, 0.40108218, 0.28643494, ..., 0.19292096, 0.2113148,
        0.07569198]])
Обучение модели и оценка ее качества
                                                                                                          In [13]:
# 2 ближайших соседа
reg1 1 = KNeighborsClassifier(n neighbors=2)
reg1 1.fit(X train, Y train)
target1_1 = reg1_1.predict(X_test)
len(target1_1), target1_1
                                                                                                         Out[13]:
(188,
array([1, 0, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0, 1, 1,
        0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1,
        1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1,
        1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1,
        0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0,
       0, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1,
       1, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 0,
        1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1,
        1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1]))
                                                                                                          In [14]:
accuracy score (Y test, target1 1)
                                                                                                         Out[14]:
0.9574468085106383
                                                                                                          In [15]:
def accuracy_score_for_classes(
    y true: np.ndarray,
    y_pred: np.ndarray) -> Dict[int, float]:
    d = {'t': y true, 'p': y pred}
    df = pd.DataFrame(data=d)
    classes = np.unique(y_true)
    res = dict()
    for c in classes:
        temp data flt = df[df['t']==c]
        temp_acc = accuracy score(
            temp data flt['t'].values,
            temp data flt['p'].values)
```

Out[11]:

```
return res
def print_accuracy_score_for_classes(
    y true: np.ndarray,
    y pred: np.ndarray):
    accs = accuracy score for classes(y true, y pred)
    if len(accs)>0:
       print('Метка \t Accuracy')
    for i in accs:
        print('{} \t {}'.format(i, accs[i]))
                                                                                                      In [16]:
# 2 ближайших соседа
print accuracy score for classes (Y test, target1 1)
Метка
      Accuracy
0 0.9538461538461539
   0.959349593495935
Кросс-валидация
                                                                                                      In [17]:
scoring = {'precision': 'precision weighted',
           'recall': 'recall weighted',
           'f1': 'f1 weighted'}
                                                                                                      In [18]:
kf = KFold(n splits=5)
scores = cross validate(KNeighborsClassifier(n neighbors=2),
                        breast cancer sc, breast cancer.target, scoring=scoring,
                        cv=kf, return train score=True)
scores
                                                                                                     Out[18]:
{'fit_time': array([0.00601196, 0.00299382, 0.00311136, 0.00397801, 0.00299215]),
 'score_time': array([0.01454091, 0.0123682 , 0.01199245, 0.0145328 , 0.00716782]),
 'test precision': array([0.95613188, 0.95010337, 0.96514312, 0.9591089, 0.93459708]),
 'train precision': array([0.98734066, 0.97166449, 0.97727737, 0.97332362, 0.97722415]),
 'test recall': array([0.95614035, 0.94736842, 0.96491228, 0.95614035, 0.92035398]),
 'train recall': array([0.98681319, 0.96923077, 0.97582418, 0.97142857, 0.97587719]),
 'test f1': array([0.95605944, 0.94756499, 0.96469636, 0.95682492, 0.92340952]),
 'train f1': array([0.98688352, 0.96948639, 0.97595915, 0.97156884, 0.9759728])}
                                                                                                      In [19]:
kf1 = LeaveOneOut()
scores1 = cross validate(KNeighborsClassifier(n neighbors=2),
                        breast cancer sc, breast cancer.target, scoring=scoring,
                        cv=kf1, return train score=True)
scores1
C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1221: UndefinedMetricWarning: Pr
ecision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter
to control this behavior.
  warn prf(average, modifier, msg start, len(result))
C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1221: UndefinedMetricWarning: Re
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```

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res[c] = temp acc

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```
_warn_prr(average, mourrier, msg_start, ren(resurt))
```

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warn prf(average, modifier, msg_start, len(result))

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_warn_prf(average, modifier, msg_start, len(result))

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warn prf(average, modifier, msg start, len(result))

C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Re call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero_division` parameter to cont rol this behavior.

warn prf(average, modifier, msg start, len(result))

C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Pr ecision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

warn prf(average, modifier, msg start, len(result))

C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Re call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero_division` parameter to cont rol this behavior.

_warn_prf(average, modifier, msg_start, len(result))

C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Pr ecision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

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C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Re call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero_division` parameter to cont rol this behavior.

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C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Re call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero_division` parameter to cont rol this behavior.

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C:\Users\enjoy\anaconda3\lib\site-\text{packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Precision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

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C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\classification.py:1221: UndefinedMetricWarning: Re

call is ill-defined and being set to U.U in labels with no true samples. Use zero division parameter to cont rol this behavior. warn prf(average, modifier, msg start, len(result)) C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Pr ecision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero division` parameter to control this behavior. warn prf(average, modifier, msg start, len(result)) $\verb|C:\Users\enjoy\anaconda3| lib\site-packages\sklearn\metrics\classification.py:1221: Undefined Metric Warning: Revenue and the state of the state$ call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero division` parameter to cont rol this behavior. warn prf(average, modifier, msg start, len(result)) C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1221: UndefinedMetricWarning: Pr ecision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero division` parameter to control this behavior. warn prf(average, modifier, msg start, len(result)) C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1221: UndefinedMetricWarning: Re call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero division` parameter to cont rol this behavior. warn prf(average, modifier, msg start, len(result)) C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Pr ecision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior. warn prf(average, modifier, msg start, len(result)) $\verb|C:|Users| enjoy| anaconda 3 lib| site-packages| sklearn| metrics| classification.py: 1221: Undefined Metric Warning: Rever the packages of the packages of$ call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero division` parameter to cont rol this behavior. warn prf(average, modifier, msg start, len(result)) C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1221: UndefinedMetricWarning: Pr ecision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero division` parameter to control this behavior. warn prf(average, modifier, msg start, len(result)) C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1221: UndefinedMetricWarning: Re call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero division` parameter to cont rol this behavior. _warn_prf(average, modifier, msg_start, len(result)) C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1221: UndefinedMetricWarning: Pr ecision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero division` parameter to control this behavior. warn prf(average, modifier, msg start, len(result)) $\verb|C:\Users\enjoy\anaconda3| lib\site-packages\sklearn\metrics\classification.py:1221: Undefined Metric Warning: Revenue and the substitution of the substitution of$ call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero division` parameter to cont rol this behavior. warn prf(average, modifier, msg start, len(result)) C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1221: UndefinedMetricWarning: Pr ecision is ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero division` parameter to control this behavior. warn prf(average, modifier, msg start, len(result)) C:\Users\enjoy\anaconda3\lib\site-packages\sklearn\metrics\ classification.py:1221: UndefinedMetricWarning: Re call is ill-defined and being set to 0.0 in labels with no true samples. Use `zero division` parameter to cont _warn_prf(average, modifier, msg_start, len(result)) Out[19]: {'fit time': array([0.00698423, 0.00178242, 0.00498557, 0.00698113, 0.00401974, , 0.01562047, 0. , 0.00502086, 0.00299239, 0.00471592, 0.00399327, 0.00402045, 0.00401902, 0.00395322, 0.00399208, 0.00299072, 0.00398946, 0.00395298, 0.01887274, 0.00495148, 0.00301766, 0.00395608, 0. , 0. , 0.01562142, 0. 0.00299239, 0. 0.01558852, 0.00335646, 0.00501466, 0.00398636, 0.00404143, 0.00398922, 0.00396013, 0.00398755, 0.00398779, 0.00299096, 0.00398684, 0.00398922, 0.00299239, 0.0039947, 0.01196647, $0.00498652,\ 0.00398874,\ 0.00598192,\ 0.00498366,\ 0.00498867,$ 0.00598288, 0.00499034, 0.00547457, 0.00598001, 0.00598216, , 0. 0.00598478, 0.01137161, 0. , 0. , 0.00498557, 0.002249 , 0.00399685, 0.00398827, , 0. , 0. , 0.0039947 , 0.00498748, 0. 0.00697923, 0.00398946, 0.01044106, 0.00402355, 0. , 0. , 0. , 0.0039885 , 0.00298786, , 0.00398469, 0.00299191, 0.01772141, 0.00299215, , 0. , 0. 0.00299287, 0. , 0.00498819, 0.00302362, 0.00401497, , 0.00234485, 0.0039885 , 0.0039897 , 0.00398827, 0. , 0.00398779, 0. , 0.00399089, 0. , 0. , 0.00600934, 0.00395966, 0.00399113, 0.

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Оптимизация гиперпараметров

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In [20]:
n range = np.array(range(1,70,2))
tuned parameters = [{'n neighbors': n range}]
tuned parameters
                                                                                                            Out[20]:
[{'n_neighbors': array([ 1, 3, 5, 7, 9, 11, 13, 15, 17, 19, 21, 23, 25, 27, 29, 31, 33,
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         69])}]
                                                                                                             In [21]:
#GridSearch
clf gs = GridSearchCV(KNeighborsClassifier(), tuned parameters, cv=kf1, scoring='accuracy')
clf_gs.fit(X_train, Y_train)
                                                                                                            Out[21]:
GridSearchCV(cv=LeaveOneOut(), estimator=KNeighborsClassifier(),
             param grid=[{'n neighbors': array([ 1, 3, 5, 7, 9, 11, 13, 15, 17, 19, 21, 23, 25, 27, 29, 31
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             scoring='accuracy')
                                                                                                             In [22]:
clf qs.cv results
                                                                                                            Out[22]:
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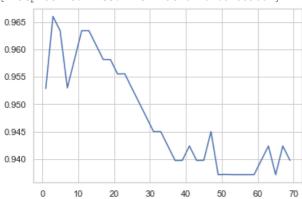
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  0.2230199 , 0.21767378, 0.2230199 , 0.2230199 , 0.2230199 ,
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  27])}
                                      In [23]:
# Лучшее значение метрики
clf gs.best score
                                     Out[23]:
0.973753280839895
                                      In [24]:
# Лучшая модель
clf gs.best estimator
                                     Out[24]:
KNeighborsClassifier(n neighbors=15)
                                      In [25]:
# Лучшее значение параметров
```

clf_gs.best_params_

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{'n neighbors': 15}
                                                                                                             In [26]:
# Изменение качества на тестовой выборке в зависимости от К-соседей
plt.plot(n range, clf gs.cv results ['mean test score'])
                                                                                                            Out[26]:
[<matplotlib.lines.Line2D at 0x2020d3c2970>]
0.975
0.970
0.965
0.960
0.955
0.950
0.945
0.940
            10
                  20
                        30
                              40
                                   50
                                         60
                                               70
#RandomizedSearchCV
clf rs = RandomizedSearchCV(KNeighborsClassifier(), tuned parameters, cv=kf1, scoring='accuracy')
clf rs.fit(X train, Y train)
                                                                                                            Out[27]:
RandomizedSearchCV(cv=LeaveOneOut(), estimator=KNeighborsClassifier(),
                   param distributions=[{'n neighbors': array([ 1, 3, 5, 7, 9, 11, 13, 15, 17, 19, 21, 23,
25, 27, 29, 31, 33,
       35, 37, 39, 41, 43, 45, 47, 49, 51, 53, 55, 57, 59, 61, 63, 65, 67,
       69])}],
                   scoring='accuracy')
                                                                                                             In [28]:
# Лучшее значение параметров
clf_rs.best_params_
                                                                                                            Out[28]:
{'n neighbors': 15}
                                                                                                             In [29]:
# Лучшая модель
clf rs.best estimator
                                                                                                            Out[29]:
KNeighborsClassifier(n_neighbors=15)
K-fold
                                                                                                             In [30]:
clf gs = GridSearchCV(KNeighborsClassifier(), tuned parameters, cv=kf, scoring='accuracy')
clf_gs.fit(X_train, Y_train)
                                                                                                            Out[30]:
GridSearchCV(cv=KFold(n splits=5, random state=None, shuffle=False),
             estimator=KNeighborsClassifier(),
             param_grid=[{'n_neighbors': array([ 1, 3, 5, 7, 9, 11, 13, 15, 17, 19, 21, 23, 25, 27, 29, 31
, 33,
       35, 37, 39, 41, 43, 45, 47, 49, 51, 53, 55, 57, 59, 61, 63, 65, 67,
       69])}],
             scoring='accuracy')
                                                                                                             In [31]:
clf_gs.best_score_, clf_gs.best_params_
                                                                                                            Out[31]:
(0.9659945317840055, {'n neighbors': 3})
                                                                                                             In [32]:
# Изменение качества на тестовой выборке в зависимости от К-соседей
plt.plot(n_range, clf_gs.cv_results_['mean_test_score'])
```

Out[25]:





clf_rs = RandomizedSearchCV(KNeighborsClassifier(), tuned_parameters, cv=kf, scoring='accuracy') clf_rs.fit(X_train, Y_train)

Out[33]:

RandomizedSearchCV(cv=KFold(n_splits=5, random_state=None, shuffle=False), estimator=KNeighborsClassifier(), param distributions=[{'n neighbors': array([1, 3, 5, 7, 9, 11, 13, 15, 17, 19, 21, 23, 25, 27, 29, 31, 33, 35, 37, 39, 41, 43, 45, 47, 49, 51, 53, 55, 57, 59, 61, 63, 65, 67, 69])}], scoring='accuracy')

clf_rs.best_score_, clf_rs.best_params_

Out[34]:

In [34]:

(0.9659945317840055, {'n_neighbors': 3})

In []: