```
from IPython.display import Image
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.datasets import load_digits
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, ShuffleS
from sklearn.metrics import accuracy score, balanced accuracy score
from sklearn.metrics import precision score, recall score, f1 score, classification report
from sklearn.metrics import confusion matrix
from sklearn.metrics import mean_absolute_error, mean_squared_error, mean_squared_log_erro
from sklearn.metrics import roc_curve, roc_auc_score
from sklearn.model_selection import GridSearchCV, RandomizedSearchCV
from sklearn.model_selection import learning_curve, validation_curve
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set(style="ticks")
mnist = load_digits()
```

▼ С использованием метода train_test_split разделите выборку на обучающ

Обучите модель ближайших соседей для произвольно заданного гиперпа модели с помощью трех подходящих для задачи метрик.

Постройте модель и оцените качество модели с использованием кросс-е эксперименты с тремя различными стратегиями кросс-валидации.

```
scores = cross_val_score(KNeighborsClassifier(n_neighbors=3),
                             mnist.data, mnist.target, cv=8)
   scores
    ray([0.94666667, 0.98222222, 0.95111111, 0.97333333, 0.96888889,
                0.99107143, 0.99107143, 0.94642857])
   np.mean(scores)
       0.9688492063492063
   scores = cross_val_score(KNeighborsClassifier(n_neighbors=3),
                             mnist.data, mnist.target, cv=8,
                            scoring='f1_weighted')
   scores, np.mean(scores)
        (array([0.94768494, 0.98200127, 0.95147097, 0.97320885, 0.96933259,
                 0.99106932, 0.9910649, 0.94565207]), 0.9689356120000322)
   scoring = {'precision': 'precision_weighted',
               'recall': 'recall_weighted',
               'f1': 'f1_weighted'}
   scores = cross_validate(KNeighborsClassifier(n_neighbors=3),
https://colab.research.google.com/drive/1vFiCrAaWf7IUdUynfO-7tdb2lLgmTfBw#scrollTo=8adBLVn8mZuU
```

▼ K-fold

```
kf = KFold(n splits=5)
scores = cross_val_score(KNeighborsClassifier(n_neighbors=3),
                         mnist.data, mnist.target, scoring='f1_weighted',
                         cv=kf)
scores
ray([0.9559596 , 0.96113277, 0.96366091, 0.98602822, 0.96628182])
kf = KFold(n_splits=5)
scores = cross_validate(KNeighborsClassifier(n_neighbors=3),
                        mnist.data, mnist.target, scoring=scoring,
                        cv=kf, return_train_score=True)
scores
    {'fit_time': array([0.01152563, 0.00672007, 0.00670028, 0.00661159, 0.00661659]),
      'score_time': array([0.07628012, 0.07173538, 0.06945038, 0.07013011, 0.07015181]),
      'test_f1': array([0.9559596 , 0.96113277, 0.96366091, 0.98602822, 0.96628182]),
      'test_precision': array([0.95762013, 0.96347273, 0.96544955, 0.9861371 , 0.96746161]
      'test_recall': array([0.95555556, 0.96111111, 0.9637883 , 0.98607242, 0.96657382]),
      'train f1': array([0.99302874, 0.99442358, 0.99233002, 0.99442686, 0.99373111]),
      'train precision': array([0.99307822, 0.99449013, 0.99238089, 0.99445488, 0.99376389
      'train recall': array([0.99304106, 0.99443285, 0.99235049, 0.99443672, 0.99374131])}
```

▼ Repeated K-fold

▼ Leave One Out (LOO)

▼ Произведите подбор гиперпараметра К с использованием GridSearchCV и

```
n_range = np.array(range(2,30,2))
tuned_parameters = [{'n_neighbors': n_range}]
tuned parameters
 \Gamma \rightarrow [\{ 'n_{neighbors}' : array([2, 4, 6, 8, 10, 12, 14, 16, 18, 20, 22, 24, 26, 28]) \}]
clf_gs = GridSearchCV(KNeighborsClassifier(), tuned_parameters, cv=5, scoring='accuracy')
clf gs.fit(mnist X train, mnist y train)
#https://stackoverflow.com/questions/49160206/does-gridsearchcv-perform-cross-validation
 GridSearchCV(cv=5, error_score=nan,
                  estimator=KNeighborsClassifier(algorithm='auto', leaf size=30,
                                                  metric='minkowski',
                                                  metric params=None, n jobs=None,
                                                  n neighbors=5, p=2,
                                                  weights='uniform'),
                  iid='deprecated', n_jobs=None,
                  param_grid=[{'n_neighbors': array([ 2, 4, 6, 8, 10, 12, 14, 16, 18, 2
                  pre dispatch='2*n jobs', refit=True, return train score=False,
                  scoring='accuracy', verbose=0)
clf_gs.cv_results_
 Гэ
```

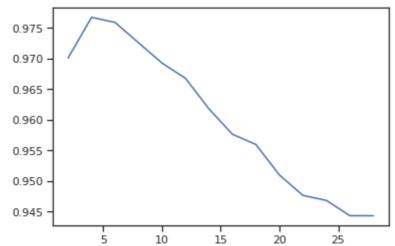
```
{'mean fit time': array([0.00546236, 0.00478535, 0.00479555, 0.00460172, 0.0045207,
             0.00463357, 0.00468388, 0.00459313, 0.004633 , 0.00458751,
             0.00515485, 0.00462193, 0.00459375, 0.00458717),
      'mean_score_time': array([0.03663001, 0.03893838, 0.03947306, 0.03825545, 0.03836875
             0.04016452, 0.03910532, 0.03924427, 0.0393961, 0.03955474,
             0.04036608, 0.03980608, 0.03997355, 0.04009004]),
      'mean_test_score': array([0.97007607, 0.97672545, 0.97589212, 0.97256916, 0.96925311
             0.96675657, 0.96177386, 0.95762102, 0.95595436, 0.95096819,
             0.94763831, 0.94681535, 0.94432227, 0.94431535]),
      'param_n_neighbors': masked_array(data=[2, 4, 6, 8, 10, 12, 14, 16, 18, 20, 22, 24,
                   mask=[False, False, False, False, False, False, False,
                         False, False, False, False, False],
             fill value='?',
                  dtype=object),
      'params': [{'n_neighbors': 2},
       {'n_neighbors': 4},
       {'n_neighbors': 6},
       {'n_neighbors': 8},
       {'n_neighbors': 10},
       {'n neighbors': 12},
       {'n_neighbors': 14},
       {'n_neighbors': 16},
       {'n_neighbors': 18},
       {'n_neighbors': 20},
       {'n_neighbors': 22},
       {'n_neighbors': 24},
       {'n_neighbors': 26},
       {'n_neighbors': 28}],
      'rank_test_score': array([ 4,  1,  2,  3,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14],
            dtype=int32),
      'split0_test_score': array([0.9626556 , 0.97095436, 0.97095436, 0.97095436, 0.958506
             0.95435685, 0.94605809, 0.93775934, 0.93775934, 0.92946058,
             0.92946058, 0.92946058, 0.92946058, 0.92946058]),
      'split1_test_score': array([0.97095436, 0.97510373, 0.97510373, 0.97095436, 0.962655
             0.96680498, 0.9626556, 0.95850622, 0.9626556, 0.95850622,
             0.9626556, 0.95435685, 0.95020747, 0.95435685]),
      'split2_test_score': array([0.97510373, 0.98340249, 0.98340249, 0.97510373, 0.975103
             0.97095436, 0.9626556, 0.95850622, 0.95435685, 0.95020747,
             0.94190871, 0.93775934, 0.93360996, 0.93775934]),
      'split3_test_score': array([0.95833333, 0.96666667, 0.9625
                                                                   , 0.9625
                                                                                , 0.966666
             0.95833333, 0.95416667, 0.95416667, 0.95
                                                          , 0.94583333,
             0.94583333, 0.94583333, 0.94166667, 0.94166667]),
      'split4 test score': array([0.98333333, 0.9875
                                                        , 0.9875
                                                                  , 0.98333333, 0.983333
                                                       , 0.97083333,
             0.98333333, 0.98333333, 0.97916667, 0.975
             0.95833333, 0.96666667, 0.96666667, 0.95833333]),
      'std_fit_time': array([1.39303721e-03, 2.10527920e-04, 2.10172868e-04, 6.98907300e-0
             8.26202632e-05, 1.31532278e-04, 1.11610111e-04, 2.96545178e-05,
             8.81757392e-05, 7.30198849e-05, 1.00808284e-03, 7.65846199e-05,
clf_gs.best_estimator_
KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
                          metric_params=None, n_jobs=None, n_neighbors=4, p=2,
                          weights='uniform')
             ע.טבעדמיט, ש.שבעבעדטי, ש.שבעדעיט, ש.שבעבעדעיט, ש.שבעבעדעיט, ש.שבעבעדעיט,
clf_gs.best_score_
C→ 0.976725449515906
```

```
clf_gs.best_params_
```

□→ {'n_neighbors': 4}

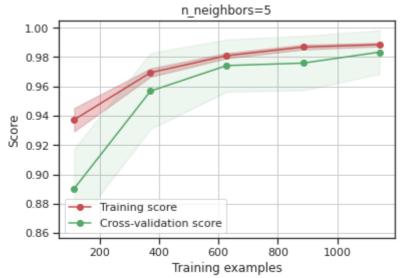
```
plt.plot(n_range, clf_gs.cv_results_['mean_test_score'])
```

[<matplotlib.lines.Line2D at 0x7f65dd248668>]



```
def plot_learning_curve(estimator, title, X, y, ylim=None, cv=None,
                        n_jobs=None, train_sizes=np.linspace(.1, 1.0, 5)):
    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="r")
    plt.fill_between(train_sizes, test_scores_mean - test_scores_std,
                     test_scores_mean + test_scores_std, alpha=0.1, color="g")
    plt.plot(train_sizes, train_scores_mean, 'o-', color="r",
             label="Training score")
    plt.plot(train_sizes, test_scores_mean, 'o-', color="g",
             label="Cross-validation score")
    plt.legend(loc="best")
    return plt
plot_learning_curve(KNeighborsClassifier(n_neighbors=5), 'n_neighbors=5',
                    mnist_X_train, mnist_y_train, cv=20)
```





Построение кривой валидации

```
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)
   1w = 2
   plt.plot(param_range, train_scores_mean, label="Training score",
                 color="darkorange", lw=lw)
   plt.fill_between(param_range, train_scores_mean - train_scores_std,
                     train_scores_mean + train_scores_std, alpha=0.4,
                     color="darkorange", lw=lw)
   plt.plot(param_range, test_scores_mean, label="Cross-validation score",
                 color="navy", lw=lw)
   plt.fill between(param range, test scores mean - test scores std,
                     test scores mean + test scores std, alpha=0.2,
                     color="navy", lw=lw)
   plt.legend(loc="best")
   return plt
n_{range2} = np.array(range(5,125,5))
```

mnist_X_train, mnist_y_train,
param_name='n_neighbors', param_range=n_range2,
cv=20, scoring="accuracy")

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