Sklearn

sklearn.liner_model

linear model:

- RidgeClassifier
- SGDClassifier
- SGDRegressor
- LinearRegression
- LogisticRegression

In [44]: **from** matplotlib.colors **import** ListedColormap

- Lasso
- etc

документация: http://scikit-learn.org/stable/modules/classes.html#module-sklearn.linear_model (http://scikit-learn.org/stable/modules/classes.html#module-sklearn.linear_model)

примеры: http://scikit-learn.org/stable/modules/linear-model (http://scikit-learn.org/stable/modules/linear-model)

```
from sklearn import cross_validation, datasets, linear_model, metrics,grid_s
import numpy as np
print(numpy.__version__)
1.11.1
```

In [45]: %pylab inline

Populating the interactive namespace from numpy and matplotlib

/Users/tatianaovchinnikova/anaconda/lib/python3.5/site-packages/IPython/c ore/magics/pylab.py:161: UserWarning: pylab import has clobbered these va riables: ['colors']

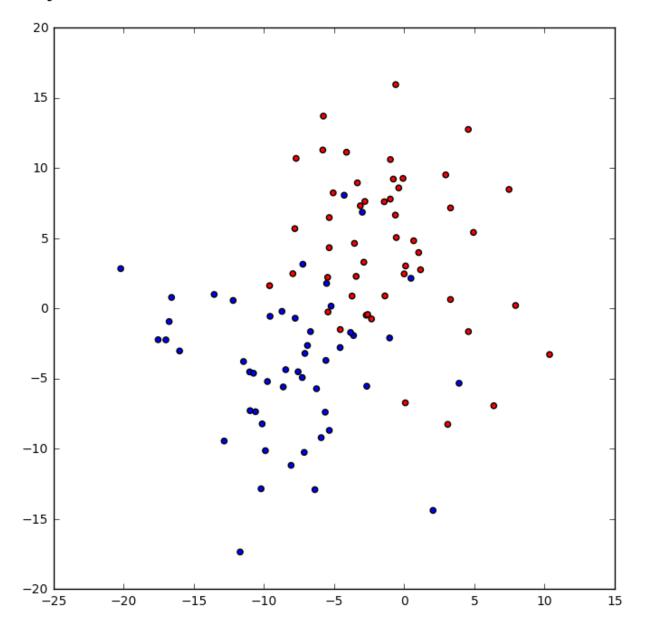
```
`%matplotlib` prevents importing * from pylab and numpy
"\n`%matplotlib` prevents importing * from pylab and numpy"
```

Генерация данных

```
In [46]: blobs = datasets.make_blobs(centers = 2, cluster_std = 5.5, random_state=1)
```

```
In [47]: colors = ListedColormap(['red', 'blue'])
    pylab.figure(figsize(8, 8))
    pylab.scatter([x[0] for x in blobs[0]], [x[1] for x in blobs[0]], c=blobs[1]
```

Out[47]: <matplotlib.collections.PathCollection at 0x10f5e6518>



In [48]: train_data, test_data, train_labels, test_labels = cross_validation.train_te

Линейная классификация

RidgeClassifier

```
In [49]:
         #создание объекта - классификатора
         ridge classifier = linear model.RidgeClassifier(random state = 1)
In [50]: #обучение классификатора
         ridge_classifier.fit(train_data, train_labels)
Out[50]: RidgeClassifier(alpha=1.0, class weight=None, copy X=True, fit intercept=
         True,
                 max_iter=None, normalize=False, random_state=1, solver='auto',
                 tol=0.001)
         #применение обученного классификатора
In [51]:
         ridge predictions = ridge classifier.predict(test data)
In [52]: print (test_labels)
         In [53]: print (ridge predictions)
         [0\ 0\ 0\ 1\ 0\ 1\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 1]
In [54]: #оценка качества классификации
         metrics.accuracy_score(test_labels, ridge_predictions)
Out[54]: 0.866666666666666
In [55]: ridge classifier.coef
Out[55]: array([[-0.0854443 , -0.07273219]])
In [56]: ridge classifier.intercept
Out[56]: array([-0.31250723])
         LogisticRegression
In [57]: log_regressor = linear_model.LogisticRegression(random_state = 1)
In [58]: log regressor.fit(train data, train labels)
Out[58]: LogisticRegression(C=1.0, class weight=None, dual=False, fit intercept=Tr
         ue,
                   intercept scaling=1, max iter=100, multi class='ovr', n jobs=1,
                   penalty='12', random state=1, solver='liblinear', tol=0.0001,
                   verbose=0, warm start=False)
In [59]: | Ir predictions = log regressor.predict(test data)
In [60]: | lr proba predictions = log regressor.predict proba(test data)
```

```
In [62]: print (test_labels)
          [0\ 0\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 1\ 1\ 1\ 0\ 0\ 0\ 1\ 1\ 0\ 0\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 0\ 1]
In [63]: print (lr_predictions)
          [0\ 1\ 1\ 1\ 0\ 1\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 1]
In [64]: print (lr proba predictions)
          [[ 0.99873457
                          0.00126543]
           [ 0.38956803
                          0.61043197]
           [ 0.45000737
                          0.549992631
           [ 0.00536816
                          0.99463184]
           [ 0.70965944
                          0.29034056]
           [ 0.23304944
                          0.76695056]
           [ 0.98983143
                          0.01016857]
           [ 0.9709671
                          0.0290329 ]
           [ 0.93909394
                          0.060906061
           [ 0.15103335
                          0.84896665]
           [ 0.90324667
                          0.096753331
           [ 0.0100448
                          0.9899552 ]
           [ 0.82338574
                          0.17661426]
           [ 0.05238427
                          0.94761573]
           [ 0.00443451
                          0.995565491
           [ 0.00102911
                          0.99897089]
                          0.791826071
           [ 0.20817393
           [ 0.91074564
                          0.08925436]
           [ 0.22580526
                          0.77419474]
           [ 0.00101709
                          0.998982911
           [ 0.09600944
                          0.903990561
                          0.00632376]
           [ 0.99367624
           [ 0.9779897
                          0.0220103 ]
                          0.70110177]
           [ 0.29889823
           [ 0.00210487
                          0.99789513]
           [ 0.00929423
                          0.990705771
           [ 0.2670277
                          0.7329723 |
           [ 0.85210662
                          0.14789338]
                          0.01128556]
           [ 0.98871444
                          0.99592428]]
           [ 0.00407572
In [65]: print (metrics.accuracy score(test labels, lr predictions))
          0.8
In [66]: print (metrics.accuracy_score(test_labels, ridge_predictions))
```

Оценка качества по cross-validation

cross_val_score

0.86666666667

```
In [67]: ridge_scoring = cross_validation.cross_val_score(ridge_classifier, blobs[0]
In [68]: | lr scoring = cross validation.cross val score(log regressor, blobs[0], blobs
In [69]: | lr scoring
                       0.9, 0.9, 0.9, 1., 1., 0.7,
Out[69]: array([ 0.7,
         print ('Ridge mean:{}, max:{}, min:{}, std:{}'.format(ridge_scoring.mean(),
                                                                ridge scoring.min(), ri
         Ridge mean: 0.8800000000000001, max: 1.0, min: 0.7, std: 0.08717797887081348
In [72]: print ('Log mean:{}, max:{}, min:{}, std:{}'.format(lr_scoring.mean(), lr_sc
                                                              lr_scoring.min(), lr_scor
         Log mean: 0.8700000000000001, max: 1.0, min: 0.7, std: 0.10049875621120892
         cross_val_score с заданными scorer и cv_strategy
         scorer = metrics.make scorer(metrics.accuracy score)
In [73]:
         cv_strategy = cross_validation.StratifiedShuffleSplit(blobs[1], n_iter = 20
In [74]:
         ridge scoring = cross validation.cross val score(ridge classifier, blobs[0]
In [75]:
         lr scoring = cross validation.cross val score(log regressor, blobs[0], blobs
         print ('Ridge mean:{}, max:{}, min:{}, std:{}'.format(ridge scoring.mean(),
                                                                ridge scoring.min(), r:
         Ridge mean: 0.8700000000000001, max: 1.0, min: 0.766666666666667, std: 0.059
         535236998305825
In [78]: print ('Log mean:{}, max:{}, min:{}, std:{}'.format(lr scoring.mean(), lr sc
                                                              lr_scoring.min(), lr_scor
         Log mean: 0.8766666666666667, max: 1.0, min: 0.7666666666666667, std: 0.06155
         395104206462
 In [ ]:
 In [ ]:
```