Carbon

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1. Load data

```
#load module data
import data
#create a dataset object and read data from file sample.txt
sample = data.DataSet()
#To read nominal data, you have to add argument 'nominal',
default is 'numeric'
#Read data from 'sample.txt'
sample.read('sample.txt', 'numeric')
#view the data
sample.x
array([[ 5.,
               1., 1., ..., 3., 1., 1.],
       [ 5.,
               4.,
                     4., ...,
                                3.,
                                      2., 1.],
              1.,
                     1., ...,
        3.,
                                3.,
                                      1.,
                                            1.1,
       5., 10., 10., ..., 8., 10.,
                                            2.1,
                                    6.,
        4.,
              8.,
                   6., ..., 10.,
                                            1.],
                     8., ...,
                                     4.,
         4.,
                8.,
                              10.,
                                            1.]])
#view class labels
sample.y[:10]
#view dataset dimension
sample.dim()
(699, 9)
#view features
sample.label
['Clump Thickness', 'Uniformity of Cell Size', 'Uniformity of
Cell Shape', 'Marginal Adhesion', 'Single Epithelial Cell Size', 'Bare Nuclei', 'Bland Chromatin', 'Normal Nucleoli', 'Mitoses']
#view subject ids
sample.key[:5]
['1000025', '1002945', '1015425', '1016277', '1017023']
#create train dataset and test dataset using 1:10 hold out
train,test = data.holdOut(sample,0.1)
```

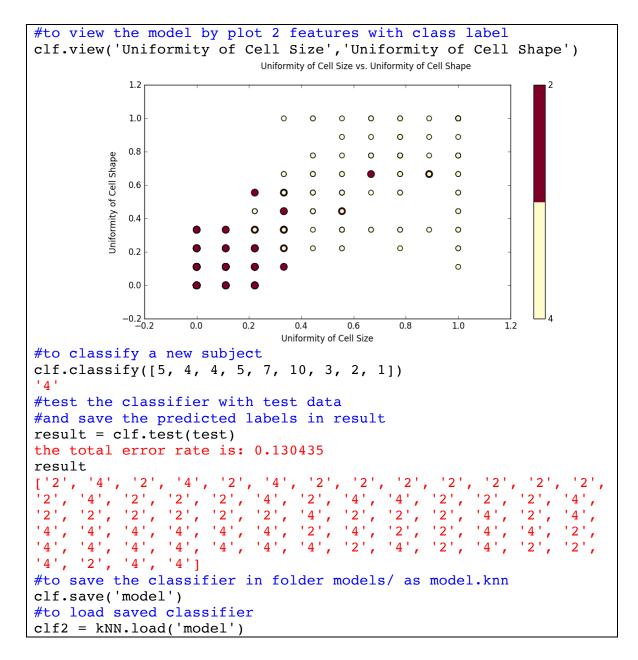
2. kNN

This is the k nearest neighbor algorithm in python.

The algorithm only works for: numerical data and nominal class Parameters:

k: int from 1 to inf distance: 'euclidean', 'correlation'

```
#load kNN module
from supervised import kNN
#create a classifier
clf = kNN.build()
#train the classifier with train data and k=4
clf.train(train,4,dist='euclidean')
```



2. ID3

ID3 is an algorithm that generate classification tree based on information gains. The algorithm only works for: nominal data and nominal class

```
#convert continuous data to nominal data
train.num2nom()
test.num2nom()
train.type
'nominal'
#load ID3 module
from supervised import ID3
#create a classifier
clf = ID3.build()
#train the classifier with train data to build the tree
```

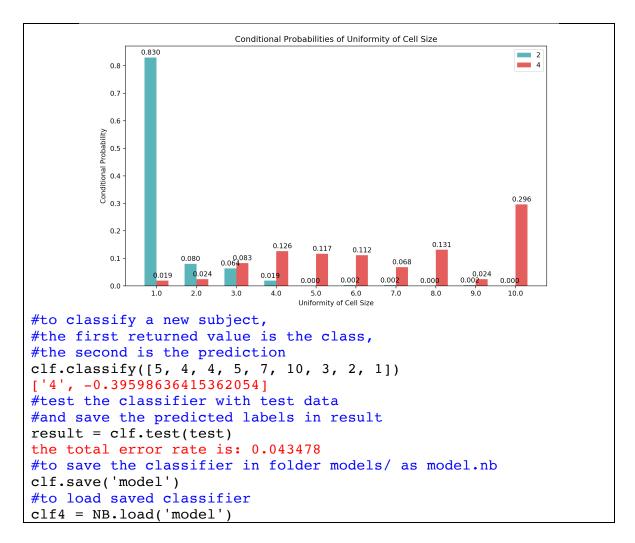
```
clf.train(train)
#to view the model
clf.view()
                             9.08.0
                                                           2.0
                                 1.0
                                     6.0
                                                           9,0 1.0 73004.0
                                   8,00300
                   10.0 3.0.0
                                                           8.0 10.0 5.2.0
                             1/6.4.0
                                    1.6.0.0
                                           8.0 10700
                                                                        8.000.8002.0
            /9/0/07/00
                                                             / a | 2 | 424244
           44424 24
                        222 4444442244444442
                                              4
                                                     42 44
                                           Clump Clump
Thicknes Thickness
                                                          Marginal Clump Clump
Adhesion Thickness Thickness
                                                                  10.0
                                                               8.5.0 4.0
                                                3.7.0 10500
                  244222
                                          44424 244442
                                                          2424 244 42
                                                                         2424
#to classify a new subject
clf.classify(['5.0', '4.0', '4.0', '5.0', '7.0', '10.0', '3.0',
'2.0', '1.0'])
'4'
#test the classifier with test data
#and save the predicted labels in result
result = clf.test(test)
the total error rate is: 0.115942
#to save the classifier in folder models/ as model.id3
clf.save('model')
#to load saved classifier
clf3 = ID3.load('model')
```

4. NB

This is the Naïve Bayes algorithm.

The algorithm only works for: nominal data and nominal class

```
#load NB module
from supervised import NB
#create a classifier
clf = NB.build()
#train the classifier with train data
clf.train(train)
#to view the model
#we can view the conditional prob of a certain feature
clf.view('Uniformity of Cell Size')
```

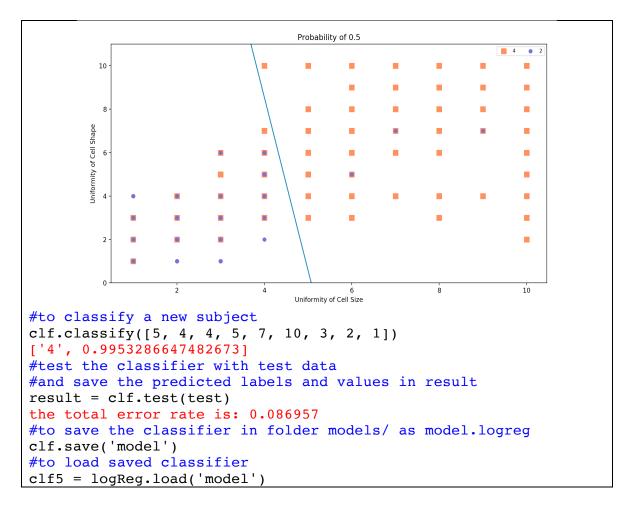


5. logReg

This is the logistic regression.

The algorithm only works for: nominal data and nominal bi-class

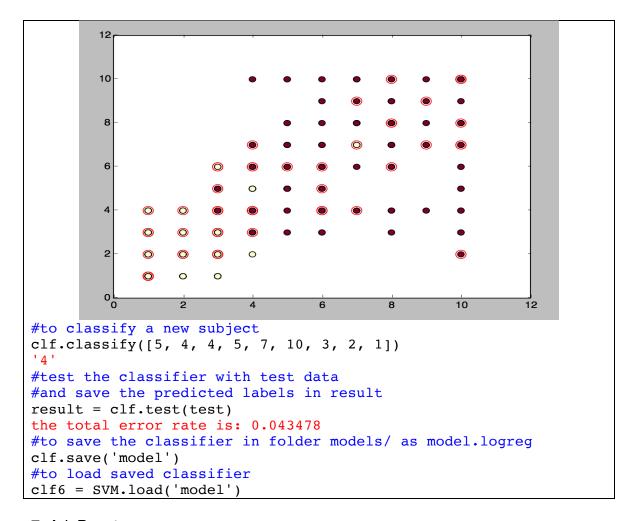
```
#load logReg module
from supervised import logReg
#create a classifier
clf = logReg.build()
#train the classifier with train data
clf.train(train)
#to view the model
#we can view the model with 2 features over train dataset
#The line in the middle is the possibility of 0.5
clf.view('Uniformity of Cell Size','Uniformity of Cell
Shape',train)
```



6. SVM

(Only works for bi-class problems)

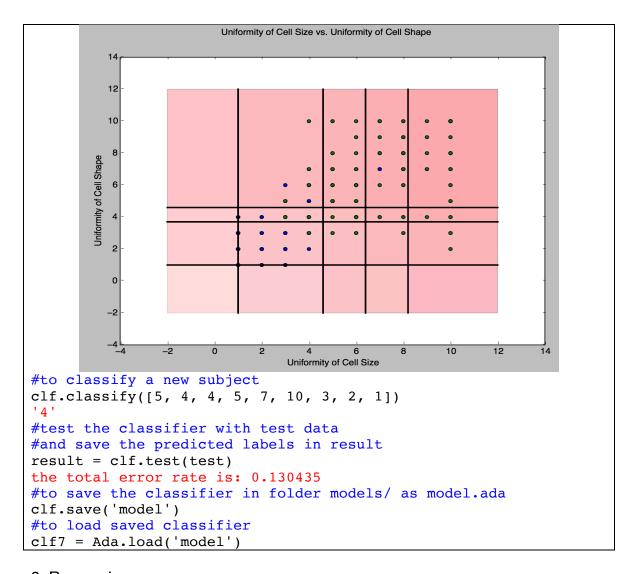
```
#load SVM module
import SVM
#create a classifier
clf = SVM.build()
#train the classifier with train data
#Here we set C to 100, tolerance to 0.001 and max iteration
number to 400. And use linear kernel.
clf.train(train,C=100,toler=0.001,maxIter=100,kTup=('lin',0))
#to view the model
#we can view the model with 2 features over train dataset
#circled points are support vectors
clf.view('Uniformity of Cell Size','Uniformity of Cell
Shape',train)
```



7. AdaBoost

(Only works for bi-class problems)

```
#load Ada module
import Ada
#create a classifier
clf = Ada.build()
#train the classifier with train data
#maximum number of iteration is 50
clf.train(train, numIt=50)
#to view the model
#we can view the model with 2 features over train dataset
#circled points are support vectors
clf.view('Uniformity of Cell Size','Uniformity of Cell
Shape',train)
```



8. Regression

There are some regression algorithms based on linear regression. We have 'linear' for regular linear regression, 'lwlr' for locally weighted linear regression.

The algorithm only works for: numerical data and numerical classes

Parameters:

```
method: 'linear', 'lwlr', 'ridge', 'lasso'
k: float
lam: float(0-1)
eps: float
numlt: int(1-inf)
```

```
#load reg module
from supervised import reg
#create a classifier
clf =reg.build()
#train the classifier with train data, and choose a method
#1.linear regression
clf.train(train, method='linear')
```

```
0.869617690359
#2.locally weighted linear regression
clf.train(train, method='lwlr', k=10)
#3.ridge linear regression
clf.train(train, method='ridge', l=100)
#4.lasso linear regression
clf.train(train, method='lasso',eps=0.01,numIt=1000)
#to view the model
#we can view the model with 1 features and class
clf.view('Uniformity of Cell Size',train)
        5
        2
#to classify a new subject
clf.classify([5, 4, 4, 5, 7, 10, 3, 2, 1])
#test the classifier with test data
#save the predicted labels in result
result = clf.test(test)
the total error rate is: 0.144928
#to save the classifier in folder models/ as model.reg
clf.save('model')
#to load saved classifier
clf8 = reg.load('model')
```

9. CART(Under Construction)

Regression binary tree building algorithm: CART.

The algorithm works for: numerical data and nominal and numeric classes Parameters:

```
model = True,False
tolS = +int
tolN = +int
```

```
#load CART module
from supervised import CART
#create a classifier
```

```
clf = CART.build()
#train the classifier with train data, minimal step 1, minimal
leaf size 4, model tree method
clf.train(train, tolS=1,tolN=4,model=False)
#to view the model as a binary tree
clf.view()
                                      Uniformity
                                        of
Cell
                                        Size
                        Uniformity
                                                      Bare
Nuclei
                         of
Cell
Size
                                        (3.89)
          (3.96)
                                                         Nuclei
                                                             >2.0
                4.00
                                                     Clump
Thickness
                                                        Uniformity
                                  (2.00)
                                              (4.00)
                            5.Shape
                                                         Shape
                                                        <=2.0 √
(2.09)
                            3.00 (3.00)
#to classify a new subject, get the class label and prediction
clf.classify([5, 4, 4, 5, 7, 10, 3, 2, 1])
('4', 4.0)
#test the classifier with test data
#save the predicted labels in result
result, values = clf.test(test)
the total error rate is: 0.173913
#to save the classifier in folder models/ as model.cart
clf.save('model')
#to load saved classifier
clf8 = reg.load('model')
```

```
#fast start
import data
sample = data.DataSet()
sample.read('sample.txt')
train,test = data.holdOut(sample,0.1)
from imp import reload

reload(CART)
clf = CART.build()
clf.train(train, tolS=1,tolN=4,model=False)
clf.classify([5, 4, 4, 5, 7, 10, 3, 2, 1])
clf.view('Uniformity of Cell Size', 'Uniformity of Cell Shape',train)
```