

test

November 4, 2022

```
[ ]: #wine
import time

import pandas as pd
import matplotlib.pyplot as plt
from pandas.plotting import radviz
from sklearn import preprocessing
from Annclass import *
# 4 1 5 RadViz
def result_visualization(test_datas, test_labels, labels_pred, accuracy, epochs):

    cols = test_labels.shape[0]
    class_true = []
    class_pred = []

    labels_true = np.argmax(test_labels, axis=1)
    #
    for i in range(cols):
        if labels_true[i] == 2:
            class_true.append(' 3')
        elif labels_true[i] == 1:
            class_true.append(' 2')
        else:
            class_true.append(' 1')
    for j in range(cols):
        if labels_pred[j] == 2:
            class_pred.append(' 3')
        elif labels_pred[j] == 1:
            class_pred.append(' 2')
        elif labels_pred[j] == 0:
            class_pred.append(' 1')
        else:
            class_pred.append(' ')

    #
    real = np.column_stack((test_datas, class_true))
    pred = np.column_stack((test_datas, class_pred))
```



```

ann = Ann(inputSize, hidSize, outputSize, learning_rate)

#
data_set_path = "data\wine_training.csv"
data_set = pd.read_csv(data_set_path, header=None)
train_datas = data_set.iloc[:, 0:inputSize].values
train_labels = data_set.iloc[:, inputSize:].values
train_datas = (train_datas - np.mean(train_datas, axis=0)) / np.
↳std(train_datas, axis=0)

#
data_test_path = "data\wine_test.csv"
data_test_set = pd.read_csv(data_test_path, header=None)
test_datas = data_test_set.iloc[:, 0:inputSize].values
test_labels = data_test_set.iloc[:, inputSize:].values
test_datas = (test_datas - np.mean(test_datas, axis=0)) / np.
↳std(test_datas, axis=0)
labels_pred = [] #
error_index = [] #
print(" ")
s = time.time()
for epoch in range(epochs):
    if (epoch + 1) % 100 == 0:
        print("    %d " % (epoch + 1))
        acc, labels_pred, labels_true, error_index = ann.
↳accuracy(test_datas, test_labels)
        print(' :', labels_pred, '\n :', labels_true, '\n :',
↳error_index)
        accuracy.append(acc)
        for i in range(len(train_datas)):
            ann.train(train_datas[i], train_labels[i])
f = time.time()
print("%d " % (epoch + 1))
print(" ", accuracy[-1])
print(" ", error_index)
print(" ", f - s, 's')
#
for i in error_index:
    #
    labels_pred[i] = -1
# 4 1 RadViz
result_visualization(test_datas, test_labels, labels_pred, accuracy, epochs)

wine_pred()

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9	0.35305020612743826	0.34867964249750577	0.6121211584417146
10	0.23885606937462805	-0.7607555836309194	0.7657917421509316
11	0.7622458628250074	-0.6815102103360319	0.20873587620502013
12	-0.4843734633931691	-1.0777370768104695	-0.6364523341956729

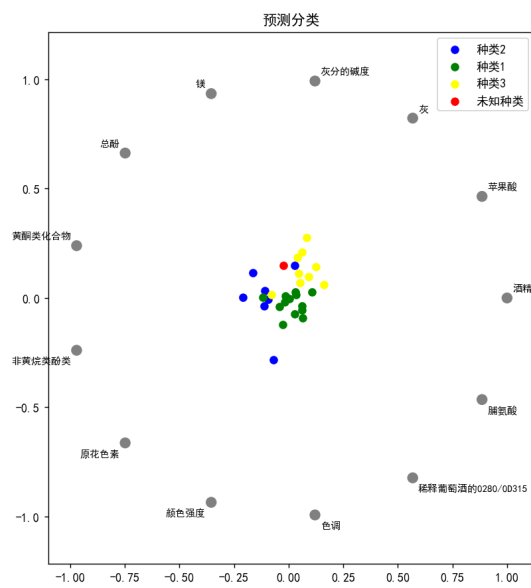
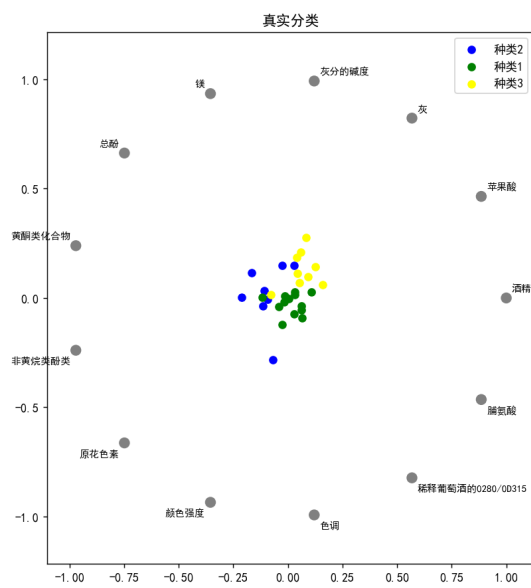
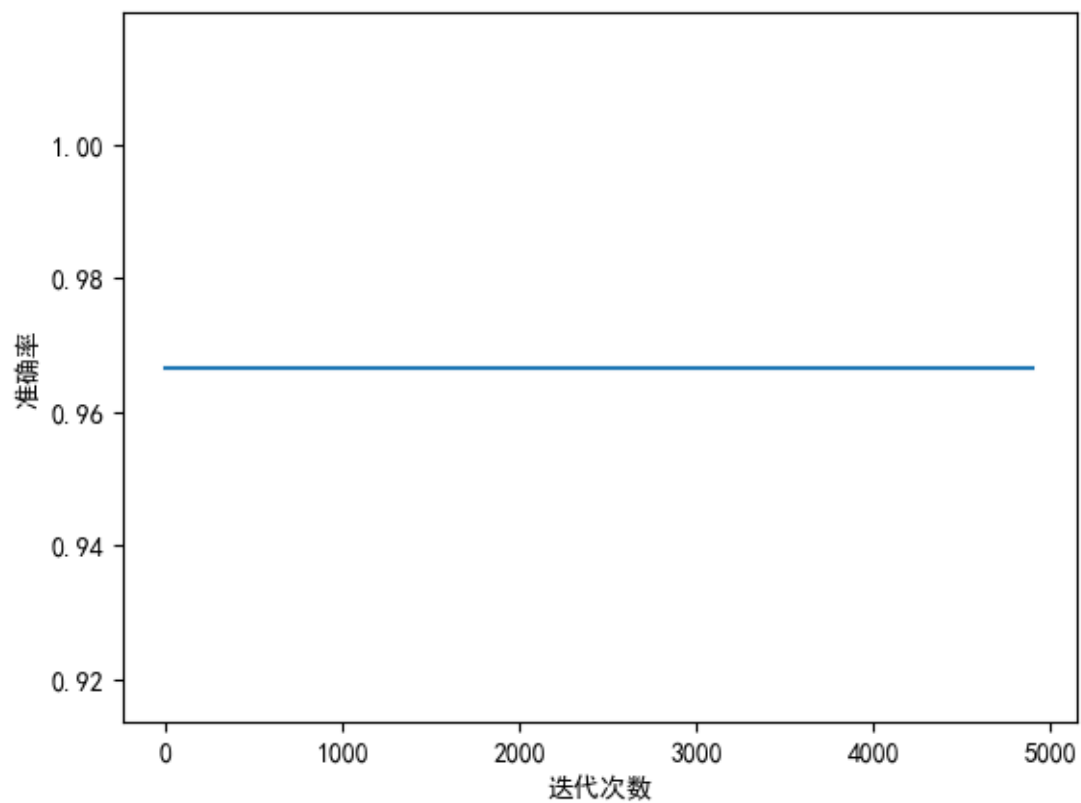
13	1.114344451146172	-0.5230194637462572	1.5341446606970162
14	-1.4074427354783838	2.171323228279918	-0.9245846786504545
15	-1.2551838864746372	0.7449065089719432	-0.6364523341956729
16	0.27692078162556477	-0.7607555836309194	-0.2330670519589785
17	-0.6556646685223844	-0.12679259727181924	-0.8477493867958463
18	0.7717620408877418	-1.1569824501053572	0.7273740962236274
19	0.3340178500019699	-0.5230194637462572	-0.27148469788628277
20	-1.3503456671019787	2.171323228279918	-1.2319258460688887
21	0.4101472745038429	-0.364528717156482	-0.6364523341956729
22	0.8574076434523492	-0.7607555836309194	0.5352858665871064
23	-0.313082258263954	0.5071703890872805	-0.44436410455915154
24	1.2475709440244505	-0.12679259727181924	0.7465829191872795
25	1.1524091633971085	-0.12679259727181924	1.4573093688424075
26	0.9716017802051594	-1.315473196695132	0.5544946895507585
27	1.2475709440244505	-1.1569824501053572	1.0731329095693658
28	0.8574076434523492	-0.5230194637462572	0.7657917421509316
29	-1.5311530502939281	1.2996241220361557	-0.8285405638321942

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0	-1.0939370559492039	2.1479536118062184	-0.4458432047918363
1	-1.3182883799748817	1.610218354330392	-0.40795284744408844
2	0.01884551121815724	-0.4062888612039585	1.2339626376249986
3	-0.35357768666446776	-0.8992128472234663	-0.3448022518645078
4	-1.0490667911440685	-0.047798689553407527	-0.6605552297624091
5	2.374534413487773	-0.5407226755729153	-1.2920611855582118
6	0.947659992684463	1.027671825398246	0.5266759671336997
7	-1.5426397040005593	0.3555027535534629	-0.9636780885443944
8	1.4681550644240353	-1.0336466615924231	-0.9889383267762265
9	-1.4663602538318288	-0.36147758974763955	0.2488133465835463
10	-0.308707421859332	0.3106914820971439	1.2718529949727466
11	-0.7125398051055519	0.6243703822913765	1.789687878725305
12	1.4412329055409538	-1.7058157334372066	-1.620444282572029
13	0.1399952261920234	0.982860553941927	0.16040251277213413
14	-0.17409662744392546	-0.675156489941872	-1.1278696370513033
15	1.3604664288917099	-1.8850608192624818	-1.2162804708627153
16	-0.6542084608588756	0.5347478393787386	1.3223734714364108
17	1.8001950239820383	-1.7954382763498444	-1.5825539252242808
18	-0.9144559967286617	1.5205958114177542	1.1581819229295023
19	-0.510623613482442	0.6691816537476954	-0.09219986954618713
20	0.7681789334639207	-1.078457933048742	-1.2668009473263797
21	-0.15166149504135776	0.4003140250097818	1.259222875856831
22	-0.12922636263879006	0.3106914820971439	1.1202915655817545
23	-1.2509829827671783	-0.8095903043108285	-0.016419154850690742
24	0.6111330066459465	0.3555027535534629	0.33722418039495905
25	0.3329373648541061	-0.09260996100972646	1.0066204935385095
26	-0.4208830838721708	1.1172943683108838	0.36248441862679115
27	0.3194762854125653	-0.2718550468350017	0.9434698979589295
28	-0.08435609783365429	0.4899365679224197	0.33722418039495905

29	0.5528016623992703	-0.8544015757671474	-1.0520889223558068
----	--------------------	---------------------	---------------------

0	-1.3900386299879974	2
1	-0.9615182067679747	2
2	0.8749978927464083	1
3	-0.6401278893529577	3
4	-0.9615182067679747	2
5	-0.3493461735965137	3
6	2.25850668771391	1
7	-0.8849967026215421	2
8	-0.18099886447436195	3
9	-1.4298298121441424	2
10	0.69134628279497	1
11	0.69134628279497	1
12	-0.9462139059386883	3
13	1.4565613242592963	1
14	-0.9003010034508286	3
15	-0.48708488106009246	3
16	0.6301290794778239	1
17	-0.7931708976458229	3
18	0.5383032745021047	2
19	0.9056064944049813	1
20	-0.3493461735965137	3
21	1.4871699259178692	1
22	0.3393473637213799	1
23	-0.7013450926701038	2
24	1.9616032516257516	1
25	-0.04326015701078322	1
26	-1.1910827192072726	2
27	0.5383032745021047	1
28	0.721954884453543	1
29	-0.8849967026215421	3



```

[ ]: #
import time

import pandas as pd
import matplotlib.pyplot as plt
from pandas.plotting import radviz
from Annclass import *

# 4 1 5 RadViz
def result_visualization(test_datas, test_labels, labels_pred, accuracy, epochs):

    cols = test_labels.shape[0]
    class_true = []
    class_pred = []

    labels_true = np.argmax(test_labels, axis=1)
    #
    for i in range(cols):
        if labels_true[i] == 2:
            class_true.append('setosa')
        elif labels_true[i] == 1:
            class_true.append('versicolor')
        else:
            class_true.append('virginica')
    for j in range(cols):
        if labels_pred[j] == 2:
            class_pred.append('setosa')
        elif labels_pred[j] == 1:
            class_pred.append('versicolor')
        elif labels_pred[j] == 0:
            class_pred.append('virginica')
        else:
            class_pred.append(' ')

    #
    real = np.column_stack((test_datas, class_true))
    pred = np.column_stack((test_datas, class_pred))

    df_real = pd.DataFrame(real, index=None,
                           columns=[' ', ' ', ' ', ' ', ' ', ' ', ' ', ' '])
    df_pred = pd.DataFrame(pred, index=None,
                           columns=[' ', ' ', ' ', ' ', ' ', ' ', ' ', ' '])

    print(df_real)
    df_real[[' ', ' ', ' ', ' ', ' ', ' ', ' ', ' ']] = df_real[
        [' ', ' ', ' ', ' ', ' ', ' ', ' ', ' ']].astype(float)

```



```

df_pred[[' ', ' ', ' ', ' ', ' ', ' ']] = df_pred[
    [' ', ' ', ' ', ' ', ' ', ' ']].astype(float)

#
#
plt.figure(" ")
plt.plot(np.arange(1, epochs, 100), accuracy)
plt.ylabel(' ')
plt.xlabel(' ')

#
plt.figure(dpi=130, figsize=(16, 8))
plt.subplot(1, 2, 1)
radviz(df_real, ' ', color=['blue', 'green', 'yellow', 'red'])
plt.title(' ')
plt.subplot(1, 2, 2)
radviz(df_pred, ' ', color=['blue', 'green', 'yellow', 'red'])
plt.title(' ')
plt.rcParams['font.sans-serif'] = ['SimHei'] #
plt.rcParams['axes.unicode_minus'] = False
plt.show()
def iris_pred():
    #
    inputSize=4          #
    hidSize = 8           #
    outputSize = 3        #
    learning_rate = 0.002  #
    epochs = 10000        #
    accuracy = []         #
    # ann
    ann = Ann(inputSize, hidSize, outputSize, learning_rate)
    #
    data_set_path = "data\iris_training.csv"
    data_set = pd.read_csv(data_set_path, header=None)
    train_datas = data_set.iloc[:, 0:inputSize].values
    train_labels = data_set.iloc[:, inputSize:].values
    #
    data_test_path = "data\iris_test.csv"
    data_test_set = pd.read_csv(data_test_path, header=None)
    test_datas = data_test_set.iloc[:, 0:inputSize].values
    test_labels = data_test_set.iloc[:, inputSize:].values
    labels_pred = [] #
    error_index = [] #
    print(" ")
    s = time.time()
    for epoch in range(epochs):
        if (epoch+1) % 100 == 0:

```

```

        print("    %d " % (epoch+1))
        acc, labels_pred, labels_true, error_index = ann.
↪accuracy(test_datas, test_labels)
        print('  :', labels_pred, '\n  :', labels_true, '\n  :',
↪error_index)
        accuracy.append(acc)
        for i in range(len(train_datas)):
            ann.train(train_datas[i], train_labels[i])
        f = time.time()
        print("%d " % (epoch+1))
        print("    ", accuracy[-1])
        print("    ", error_index)
        print("    ", f-s, 's')
        #
        for i in error_index:
            #
            labels_pred[i]=-1
        # 4      1      RadViz
        result_visualization(test_datas, test_labels, labels_pred, accuracy, epochs)
iris_pred()

```

```

100
: [2 2 1 1 0 1 2 1 1 1 2 1 1 1 2 1 2 1 2 2 2 0 0 2 1 0 0 0 1 1]
: [2 2 1 1 0 1 2 1 1 1 2 1 1 2 2 1 2 1 2 2 2 0 0 2 1 0 0 0 1 1]
: [13]
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: [2 2 1 1 0 1 2 1 1 1 2 1 1 2 2 1 2 1 2 2 2 0 0 2 1 0 0 0 1 1]
: [ 2  5  7  8  9 11 12 15 17 24 29]
300
: [2 2 1 1 0 1 2 1 1 1 2 1 1 1 2 1 2 1 2 2 2 0 0 2 1 0 0 0 1 1]
: [2 2 1 1 0 1 2 1 1 1 2 1 1 2 2 1 2 1 2 2 2 0 0 2 1 0 0 0 1 1]
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: [2 2 1 1 0 1 2 1 1 1 2 1 1 2 2 1 2 1 2 2 2 0 0 2 1 0 0 0 1 1]
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: [ 0  1 13 16]
700

```

```

: [2 2 1 1 0 1 2 1 1 1 2 2 1 1 2 1 2 1 2 2 2 0 0 2 1 0 0 0 1 1]
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1300
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: [11 13]
1900

```



```

: [2 2 1 1 0 1 2 1 1 1 2 1 1 1 2 1 2 1 2 2 2 0 0 2 1 0 0 0 1 1]
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9700
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9800
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: [2 2 1 1 0 1 2 1 1 1 2 1 1 2 2 1 2 1 2 2 2 0 0 2 1 0 0 0 1 1]
: [13]
10000
0.9666666666666667
[13]
33.68721961975098 s

0  7.7  3.0  6.1  2.3      setosa
1  7.4  2.8  6.1  1.9      setosa
2  6.8  2.8  4.8  1.4  versicolor
3  6.7  3.1  4.4  1.4  versicolor

```

4	5.5	3.5	1.3	0.2	virginica
5	5.6	3.0	4.5	1.5	versicolor
6	6.4	2.8	5.6	2.1	setosa
7	5.8	2.7	3.9	1.2	versicolor
8	6.9	3.1	4.9	1.5	versicolor
9	6.4	2.9	4.3	1.3	versicolor
10	6.4	2.8	5.6	2.2	setosa
11	6.2	2.2	4.5	1.5	versicolor
12	5.0	2.0	3.5	1.0	versicolor
13	7.9	3.8	6.4	2.0	setosa
14	7.6	3.0	6.6	2.1	setosa
15	5.7	2.9	4.2	1.3	versicolor
16	6.4	2.7	5.3	1.9	setosa
17	5.5	2.6	4.4	1.2	versicolor
18	5.8	2.7	5.1	1.9	setosa
19	4.9	2.5	4.5	1.7	setosa
20	6.7	3.3	5.7	2.5	setosa
21	4.7	3.2	1.6	0.2	virginica
22	5.7	3.8	1.7	0.3	virginica
23	5.7	2.5	5.0	2.0	setosa
24	6.5	2.8	4.6	1.5	versicolor
25	5.3	3.7	1.5	0.2	virginica
26	5.1	3.7	1.5	0.4	virginica
27	5.0	3.0	1.6	0.2	virginica
28	6.1	2.8	4.0	1.3	versicolor
29	6.0	3.4	4.5	1.6	versicolor

