

Phone accelerometer

Documentatie

Suport vector machine - Razvan Cazacu

Pentru clasificarea utilizatorilor de telefoane am ales sa folosesc un clasificator SVM cu un **parametru de penalitate** cu valoarea **C=9,1**.

Parametrul de penalitate a fost ales dupa mai multe incercari pentru a evita overfitting, respectiv underfitting.

Kenrel-ul ales pentru SVM este **RBF** cu **gamma='scale'**.

Kernel RBF:

$$K(u, v) = \exp(-\gamma * \|u - v\|^2)$$

Datele primite pentru fiecare persoana sunt inregistrate pe parcursul a 1,5 secunde la o frecventa de 100 Hz, astfel rezultant intr-un nr de valori de la 136 pana la 159.

Pentru ca datele au fost inregistrate in acest mod, o modalitate eficienta de a le prelucra ar fi fost aplicarea unui FFT (Fast Fourier transform).

Pentru a antrena SVM-ul am folosit mai multe metode, iar pentru cazul prezentat aici, am folosit raw data la care am adaugat un feature [average(col_x), average(col_y), average(col_z)] pentru fiecare linie care nu avea suficiente feature-uri pana la 150, iar pentru cele care depaseau 150, nu am utilizat datele respective.

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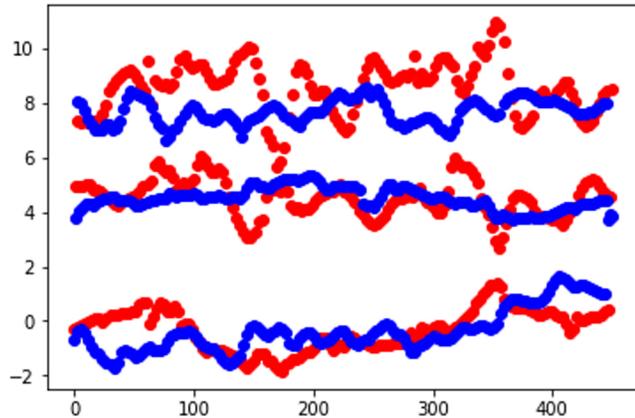
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Code

Am folosit pandas si numpy pentru a stoca datele date :

	0	1							
0	id	class							
1	10003	7		[-0.767342 8.673594]	4.330513	7.996035	...	0.52852	4.182671
2	10005	13		[-1.105523 3.93028133]	4.081836	8.904634	...	3.943649	3.93926
3	10006	6		[-0.226252 7.870339]	3.438672	8.314464	...	-2.042853	4.103063
4	10007	1							

Labels training data



Ilustrarea datelor pentru 2 utilizatori (x, y, z)

Training data

```
train_data_lines = []
for root, dirs, files in os.walk(dataPath + "train"):
    for x in files:
        train_data = np.array(pd.read_csv(dataPath
        avgx = train_data[:,0].mean()
        avgy = train_data[:,1].mean()
        avgz = train_data[:,2].mean()
        for i in range(train_data.shape[0],150):
            train_data = np.concatenate(
                (train_data,
                 [[avgx,avgy,avgz]])
            ),axis=0)
        train_data = train_data.flatten()
        train_data_lines.append(train_data[:450])
train_data_lines = np.array(train_data_lines)
plt.plot(train_data_lines[0],"ro")
plt.plot(train_data_lines[1],"bo")
```

Codul respectiv

Am prelucrat datele folosind StandardScaler pentru restrangerea datelor intr-un interval si a imbunatatit acuratetea, iar dupa aceea am antrenat SVM-ul pe datele de training (9000, 450) si am dat predict pe cele de test (5000, 450).

Pentru a nu lucra orbeste si sa nu fac overfitting, mi-am impartit datele de antrenare in 80% training si 20% test pentru a observa comportamentul SVM-ului.

Am folosit si KFold pentru a imparti datele de antrenare, dar in final am ramas la a utiliza 80% - 20%

```
from sklearn.model_selection import KFold
kf = KFold(5,True,1)
# kf.get_n_splits(scaled_train_data)
#cross validation
for train_index, test_index in kf.split(scaled_train_data):
    print("TRAIN:", train_index, "TEST:", test_index)
    X_train, X_test = scaled_train_data[train_index], scaled_train_data[test_index]
    y_train, y_test = np_train_labels[train_index,[1]], np_train_labels[test_index,[1]]
```

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```
[[ 70 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 6 0 0 0]
[ 0 96 0 0 0 0 0 0 0 0 2 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 89 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 65 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 0 79 0 0 0 5 0 3 0 1 3 0 3 0 0 0 0 0 3]
[ 0 0 0 0 0 89 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0]
[ 0 1 0 0 0 0 75 0 0 0 0 2 0 0 0 0 0 1 0 0 0 0]
[ 0 0 0 0 0 0 86 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 0 2 0 0 0 85 0 1 0 0 1 0 0 0 0 0 0 0]
[ 0 0 0 0 0 0 0 0 0 96 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 0 2 0 0 0 6 0 81 0 0 5 0 0 0 0 0 0 1]
[ 0 8 0 0 0 0 3 0 0 0 0 0 72 0 0 3 0 0 0 0 0 0]
[ 0 0 0 0 0 0 0 0 0 0 2 84 0 0 0 0 0 0 0 0 0]
[ 0 0 0 0 4 0 0 0 0 0 4 0 0 90 0 1 0 0 0 0 3]
[ 0 2 0 0 0 0 5 0 0 0 0 6 1 0 76 0 0 0 0 0 0]
[ 0 0 0 0 5 0 0 0 0 0 1 0 1 1 0 99 0 0 0 4]
[ 0 1 0 0 0 0 1 0 0 0 0 1 0 0 1 0 76 0 0 0]
[ 3 0 0 0 0 0 2 0 0 1 0 0 0 0 0 0 0 0 1 102 0 0]
[ 0 0 0 0 6 0 0 0 0 0 0 0 0 0 0 0 0 0 0 87 0]
[ 0 0 0 3 0 0 0 0 0 0 0 1 0 6 0 0 0 0 71]]
```

9 | train acc = 0.9981944444444445 | test acc = 0.926666666666666

```
[[ 70 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 6 0 0 0]
[ 0 96 0 0 0 0 0 0 0 0 2 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 89 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 65 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 0 79 0 0 0 5 0 3 0 1 3 0 3 0 0 0 0 0 3]
[ 0 0 0 0 0 89 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0]
[ 0 1 0 0 0 0 75 0 0 0 2 0 0 0 0 0 1 0 0 0 0]
[ 0 0 0 0 0 0 86 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 0 2 0 0 0 85 0 1 0 0 1 0 0 0 0 0 0 0]
[ 0 0 0 0 0 0 0 0 96 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 0 0 0 0 2 0 0 0 6 0 81 0 0 5 0 0 0 0 0 0 1]
[ 0 8 0 0 0 0 3 0 0 0 0 0 72 0 0 3 0 0 0 0 0 0]
[ 0 0 0 0 0 0 0 0 0 0 2 84 0 0 0 0 0 0 0 0 0]
[ 0 0 0 0 4 0 0 0 0 0 4 0 0 90 0 1 0 0 0 0 3]
[ 0 2 0 0 0 0 5 0 0 0 0 6 1 0 76 0 0 0 0 0 0]
[ 0 0 0 0 5 0 0 0 0 0 1 0 1 1 0 99 0 0 0 4]
[ 0 1 0 0 0 0 1 0 0 0 0 1 0 0 0 0 0 0 0 77 0 0]
[ 3 0 0 0 0 0 2 0 0 1 0 0 0 0 0 0 0 0 0 1 102 0 0]
[ 0 0 0 0 6 0 0 0 0 0 0 0 0 0 0 0 0 0 0 87 0]
[ 0 0 0 3 0 0 0 0 0 0 0 1 0 6 0 0 0 0 71]]
```

9.1 | train acc = 0.998333333333333 | test acc = 0.927222222222222

[0.9981944444444445, 0.9981944444444445, 0.998333333333333, 0.998333333333333, 0.998333333333333]

Pentru a observa comportamentul diferit am mai folosit date prelucrate diferit:
raw data umplute cu 0-uri, cu media facuta pe ultimele 2 elemente, am incercat
sa imi creez mai multe feature-uri folosind functii matematice de statistica aplicate
pe coloanele x, y, z (ex: mean, stdev, variance etc.).