

## [matlab project] svm, rf 분류기 만들기

### 1.데이터 불러오기

```
T=readtable('/MATLAB Drive/matlab 공모전/TTdata.csv');  
f1 = T{:, 'reye_ratio'}
```

```
f1 = 14630x1  
    0.2775  
    0.2775  
    0.2775  
    0.2775  
    0.2630  
    0.2891  
    0.2775  
    0.2936  
    0.2775  
    0.3084  
     ...  
     ...  
     ...
```

```
f2 = T{:, 'leye_ratio'}
```

```
f2 = 14630x1  
    0.2911  
    0.2936  
    0.2911  
    0.2936  
    0.2911  
    0.3049  
    0.2775  
    0.2936  
    0.2951  
    0.3039  
     ...  
     ...  
     ...
```

```
f3 = T{:, 'eye_angle'}
```

```
f3 = 14630x1  
   -0.9596  
   -0.9742  
   -0.9596  
   -0.9742  
   -0.9612  
   -0.9769  
   -0.9756  
   -0.9727  
   -0.9857  
   -0.9920  
     ...  
     ...  
     ...
```

```
f4 = T{:, 'nose_ratio'}
```

```
f4 = 14630x1  
    0.4877  
    0.4877  
    0.4877  
    0.4877  
    0.4877  
    0.5122
```

```
0.4877
0.4877
0.4877
0.4887
⋮
⋮
```

```
f5 = T[:, 'chin_sharp']
```

```
f5 = 14630x1
-0.3264
-0.3264
-0.3628
-0.3246
-0.3821
-0.3773
-0.3897
-0.3246
-0.3313
-0.3518
⋮
⋮
```

<<<중요! 원핫인코딩 라벨 불러오기>>>softmax는 원핫.

```
X= [f1,f2,f3,f4,f5]
```

```
X = 14630x5
0.2775    0.2911   -0.9596    0.4877   -0.3264
0.2775    0.2936   -0.9742    0.4877   -0.3264
0.2775    0.2911   -0.9596    0.4877   -0.3628
0.2775    0.2936   -0.9742    0.4877   -0.3246
0.2630    0.2911   -0.9612    0.4877   -0.3821
0.2891    0.3049   -0.9769    0.5122   -0.3773
0.2775    0.2775   -0.9756    0.4877   -0.3897
0.2936    0.2936   -0.9727    0.4877   -0.3246
0.2775    0.2951   -0.9857    0.4877   -0.3313
0.3084    0.3039   -0.9920    0.4887   -0.3518
⋮
⋮
```

```
y=T[:, 'tal']
```

```
y = 14630x1
0
0
0
0
0
0
0
0
0
0
0
⋮
⋮
```

```
save('TTdata.mat', 'X', 'y')
```

## 2. 데이터 스플릿

```
%training set, 0% validation set and 30% test set.
```

```
[train_idx, ~, test_idx] = dividerand(14630, 0.8, 0.0, 0.2);
x_train = X(train_idx, :);
y_train = y(train_idx);
x_test = X(test_idx, :);
y_test = y(test_idx);
```

### 3. SVM ECOC 모델 만들기

```
t = templateSVM('Standardize', true)
```

t =  
피팅할 템플릿 대상 classification SVM.

```
Alpha: [0x1 double]
BoxConstraint: []
CacheSize: []
CachingMethod: ''
ClipAlphas: []
DeltaGradientTolerance: []
Epsilon: []
GapTolerance: []
KKTolerance: []
IterationLimit: []
KernelFunction: ''
KernelScale: []
KernelOffset: []
KernelPolynomialOrder: []
NumPrint: []
Nu: []
OutlierFraction: []
RemoveDuplicates: []
ShrinkagePeriod: []
Solver: ''
StandardizeData: 1
SaveSupportVectors: []
VerbosityLevel: []
Version: 2
Method: 'SVM'
Type: 'classification'
```

```
svm_ecoc = fitcecoc(x_train, y_train);
```

### 4. SVM-ECOC의 클래스는 0-3, 커널 파라미터는 struct값 그대로이다.

```
talname = svm_ecoc.ClassNames
```

```
talname = 4x1
0
1
2
3
```

### svm -ecoc 모델 하이퍼파라미터

```
svm_ecoc.BinaryLearners{1}
```

```
ans =
CompactClassificationSVM
ResponseName: 'Y'
CategoricalPredictors: []
```

```

        ClassNames: [-1 1]
        ScoreTransform: 'none'
            Beta: [5×1 double]
            Bias: 6.5308
        KernelParameters: [1×1 struct]

```

Properties, Methods

## svm-ecoc 모델 로스 결과

```
error = resubLoss(svm_ecoc)
```

```
error = 0.2434
```

## 4 SVM ECOC 교차검증

```
CVsvm_1 = crossval(svm_ecoc);
```

## 5. 교차검증 확인하기

```
genError = kfoldLoss(CVsvm_1)
```

```
genError = 0.2440
```

## 6. SVM\_ECOC 2926장 테스트셋(0.2) 예측

```
pred_svm_ecoc = svm_ecoc.predict(x_test)
```

```

pred_svm_ecoc = 2926×1
    0
    0
    0
    0
    0
    3
    3
    3
    3
    3
    ⋮
    ⋮

```

## 7. SVM\_ECOC 로스와 오차행렬

```
svmecocResubErr = resubLoss(svm_ecoc)
```

```
svmecocResubErr = 0.2434
```

```
svm_ecocCM = confusionchart(y_test,pred_svm_ecoc);
```

|   |        |     |     |     |
|---|--------|-----|-----|-----|
| 0 | 588    | 48  | 8   | 82  |
| 1 | 54     | 596 | 85  | 21  |
| 2 | 10     | 112 | 601 | 100 |
| 3 | 70     | 15  | 124 | 412 |
|   | 0      | 1   | 2   | 3   |
|   | 예측 클래스 |     |     |     |

[8. SVM cidscr] : 훈련데이터에 대한 오분류. 테스트 데이터 predict 없음.

```
%ficdiscr 함수 모델링 결과 : 아래와 같다.
```

```
figure
```

```
svm_c = fitcdiscr(x_train,y_train)
```

```
svm_c =  
ClassificationDiscriminant  
    ResponseName: 'Y'  
    CategoricalPredictors: []  
    ClassNames: [0 1 2 3]  
    ScoreTransform: 'none'  
    NumObservations: 11704  
    DiscrimType: 'linear'  
    Mu: [4x5 double]  
    Coeffs: [4x4 struct]
```

Properties, Methods

```
pred_svm_c = resubPredict(svm_c)
```

```
pred_svm_c = 11704x1
```

```
0  
0  
0  
0  
0  
0  
3  
3  
3  
0  
:  
:
```

```
svm_cCM = confusionchart(y_train,pred_svm_c);
```

|   |        |      |      |      |
|---|--------|------|------|------|
| 0 | 2308   | 194  | 55   | 292  |
| 1 | 215    | 2509 | 328  | 64   |
| 2 | 55     | 278  | 2762 | 306  |
| 3 | 303    | 83   | 453  | 1499 |
|   | 0      | 1    | 2    | 3    |
|   | 예측 클래스 |      |      |      |

```
svm_cResubErr = resubLoss(svm_c)
```

```
svm_cResubErr = 0.2244
```

## 6. SVM -ceco 최적화하기(c는 교차검증할 데이터, 훈련데이터의 일부이다)

```
%rng default  
%svm_op = fitcecoc(x_train,y_train,'OptimizeHyperparameters','auto',...  
%     'HyperparameterOptimizationOptions',struct('AcquisitionFunctionName',...  
%     'expected-improvement-plus'))
```