TU: Decision Tree

Industrial AI & Automation by Y.K.Kim

Mod: 2024-2

Author: Jin Kwak/21900031

Date: 24.09.20

Introduction

Classification with decision tree

Example

Dataset: CWRU dataset features

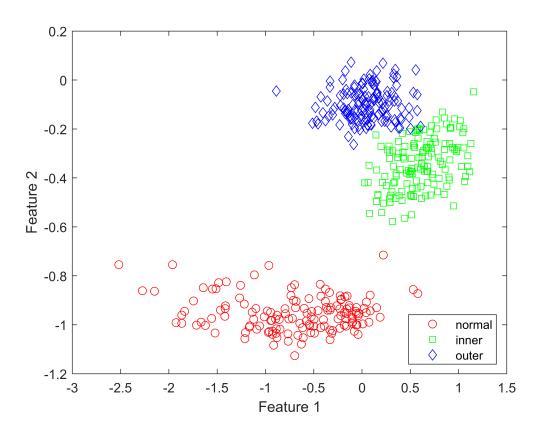
- Given dataset contains many features extracted from CWRU dataset
- We will select 2~3 features for exercise

```
clear all;
%% Train
addpath("../../Dataset/CWRU_selected_Dataset/Feature_data/");
load("sample_train.mat");
feature1 = "sv";
                                % skewness value of time data
feature2 = "ipf";
                                % impulse factor
X(:, 1) = table2array(glob_all_train(:, feature1));
X(:, 2) = table2array(glob_all_train(:, feature2));
Y = class cwru train;
                      % fault class
N = size(X,1);
tbl=table(X(:, 1),X(:, 2),Y);
load("../../Dataset/CWRU_selected_dataset/Feature_data/sample_test.mat");
Xtest(:, 1) = table2array(glob_all_test(:, feature1));
Xtest(:, 2) = table2array(glob_all_test(:, feature2));
Ytest = class_cwru_test;
Ntest=size(Xtest,1);
tblTest=table(Xtest(:, 1), Xtest(:, 2), Ytest);
```

Plot Test Data

```
f = figure;
gscatter(X(:, 1), X(:, 2), Y,'rgb','osd');
xlabel('Feature 1')
```

ylabel('Feature 2')



Prepare Cross-Validation Data

cvpartition to generate 10 disjoint stratified subsets.

Fit decision Tree

```
t = fitctree(X, Y, 'PredictorNames', {'SV' 'IF' });
```

Anlayze (Train Data)

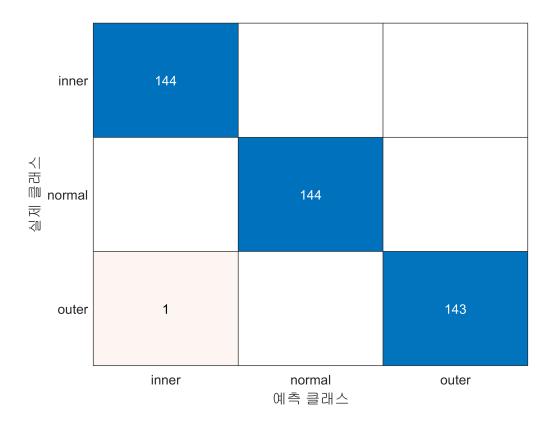
Resubstitution error: misclassification error on the training set.

```
test_resub = resubLoss(t)
```

```
test_resub = 0.0023
```

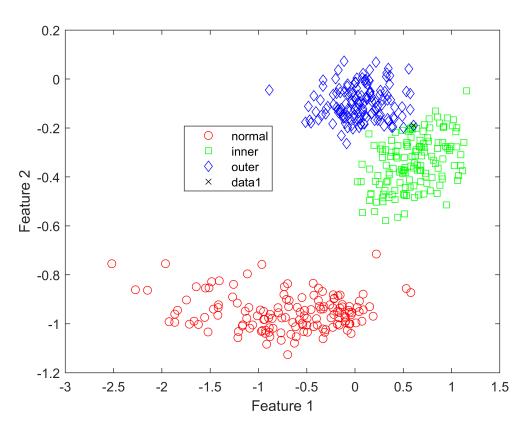
Confusion matrix on the training set

```
test_class = resubPredict(t);
figure
tResubCM = confusionchart(Y,test_class);
```

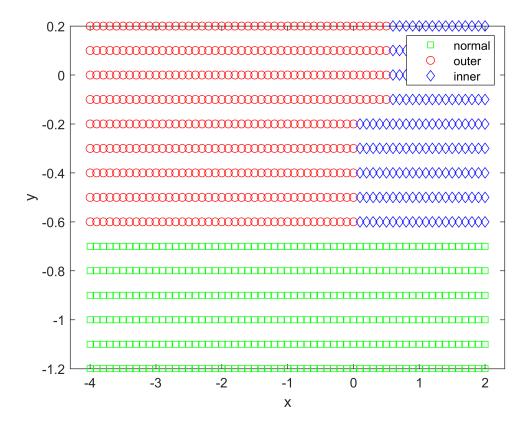


Misclassification Plot

```
figure(f)
bad = ~strcmp(test_class,Y);
hold on;
plot(X(bad,1), X(bad,2), 'kx');
hold off;
```



```
% Visualize Separation regions
[x,y] = meshgrid(-4:.1:2,-1.2:.1:0.2);
x = x(:);
y = y(:);
[grpname,node] = predict(t,[x y]);
gscatter(x,y,grpname,'grb','sod')
```



Validation (CV)

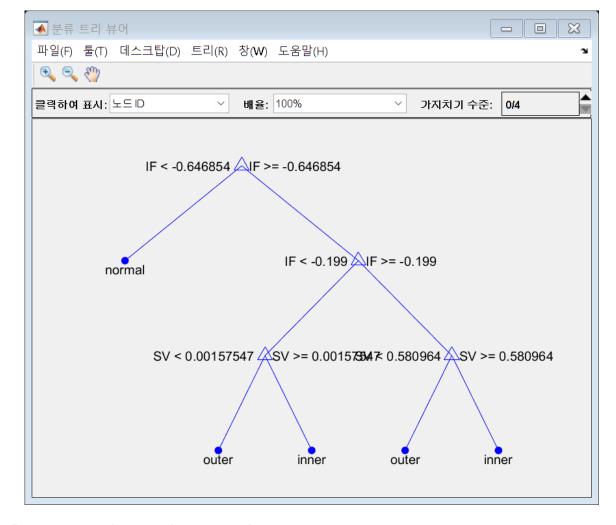
Estimate the true test error for LDA using 10-fold stratified cross-validation.

```
cvt = crossval(t,'CVPartition',cv);
tCVErr = kfoldLoss(cvt)

tCVErr = 0.0069
```

visualize the decision tree

```
view(t,'Mode','graph');
```



Prune tree nodes to enhance result

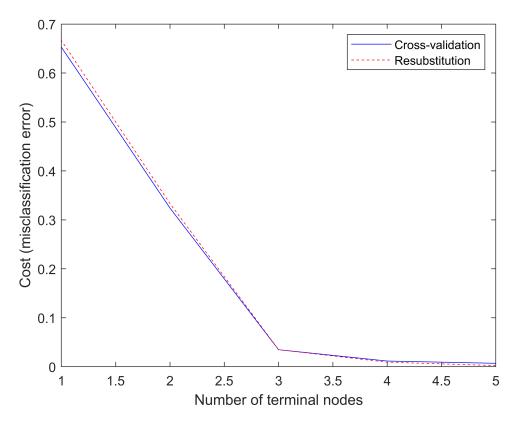
- First compute the resubstitution error for various subsets of the original tree.
- Then compute the cross-validation error for these sub-trees.

```
resub_loss = resubLoss(t,'Subtrees','all')

resub_loss = 5×1
    0.0023
    0.0093
    0.0347
    0.3333
    0.6667

[cost,se_cost,n_leaf,bestlevel] = cvloss(t,'Subtrees','all');

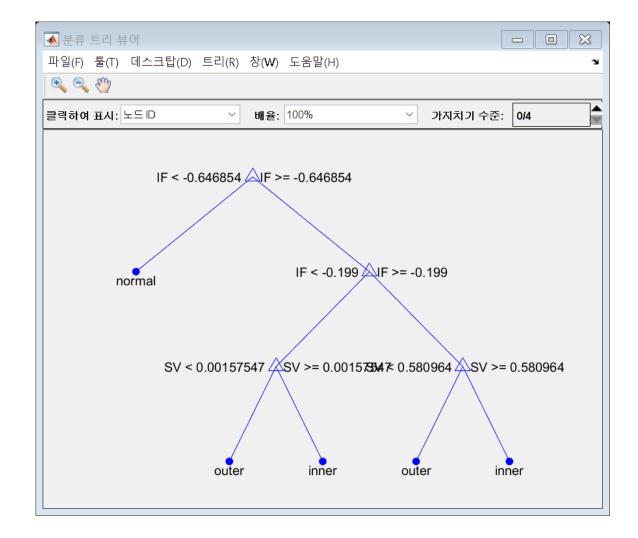
plot(n_leaf,cost,'b-', n_leaf,resub_loss,'r--')
xlabel('Number of terminal nodes');
ylabel('Cost (misclassification error)')
legend('Cross-validation','Resubstitution')
```



```
% Prune Tree Levels
pt = prune(t,'Level',bestlevel);
cost(bestlevel+1)
```

ans = 0.0069

view(pt,'Mode','graph')



Exercise

Exercise 1:

Now, Select any 3 features and repeat the process.

Features

Plot the test and train results in 3D graph.

```
fig1 = figure;

tableX = glob_all_train(:,{'kv','mf','cf'});
hold on;
scatter3(tableX,'kv','mf','cf','ColorVariable','kv');
xlabel('Feature 1: Kurtosis Value');
```

```
ylabel('Feature 2: Marginal-Factor');
zlabel('Feature 3: Marginal-Factor');
```

```
0.2
       0
Feature 2: Marginal-Factor
    -0.2
     -0.4
    -0.6
    -0.8
    -1.2
        -0.9
                             -0.7
                                       -0.6
                                                  -0.5
                                                             -0.4
                                                                                 -0.2
                                                                                            -0.1
                  -0.8
                                                                       -0.3
                                        Feature 1: Kurtosis Value
```

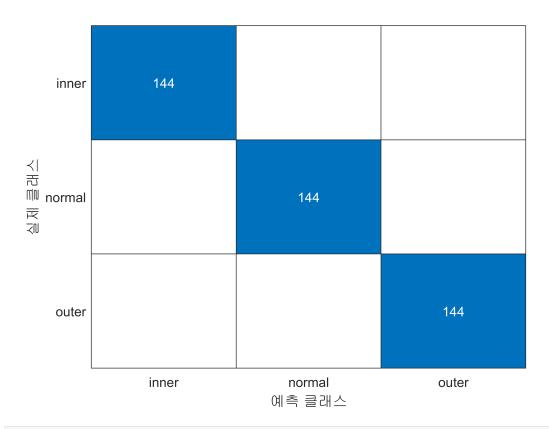
```
Y = class_cwru_train; % fault class
N = size(X,1);
tbl=table(X(:, 1),X(:, 2),X(:, 3),Y);
```

```
Xtest(:, 1) = table2array(glob_all_test(:, "kv"));
Xtest(:, 2) = table2array(glob_all_test(:, "mf"));
Xtest(:, 3) = table2array(glob_all_test(:, "cf"));
Ytest = class_cwru_test;
Ntest=size(Xtest,1);
tblTest=table(Xtest(:, 1),Xtest(:, 2),Xtest(:, 3),Ytest);
```

```
rng(0)
cv = cvpartition(Y,'KFold',10)
```

```
t = fitctree(X, Y, 'PredictorNames', {'sv' 'ipf' 'cf' });
test_resub = resubLoss(t);
test_class = resubPredict(t);
```

```
figure;
tResubCM = confusionchart(Y,test_class);
```



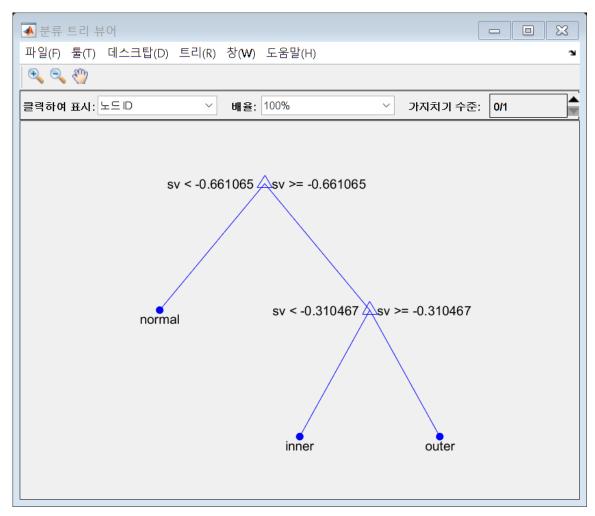
```
bad = ~strcmp(test_class,Y);
figure(fig1); hold on;
```

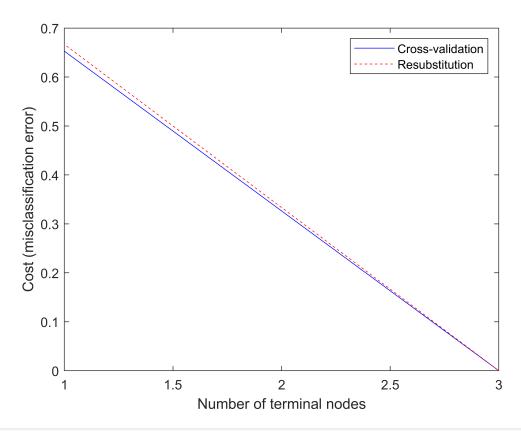
```
0
Feature 2: Marginal-Factor
    -0.2
    -0.4
    -0.6
    -0.8
    -1.2 -0.9
                                                 -0.5
                  -0.8
                            -0.7
                                       -0.6
                                                           -0.4
                                                                      -0.3
                                                                                -0.2
                                                                                          -0.1
                                                                                                       0
                                       Feature 1: Kurtosis Value
```

0.2

```
plot(X(bad,1), X(bad,2), 'kp');
```

```
cross_val = crossval(t,'CVPartition',cv);
test_cv_error = kfoldLoss(cross_val);
view(t,'Mode','graph');
```

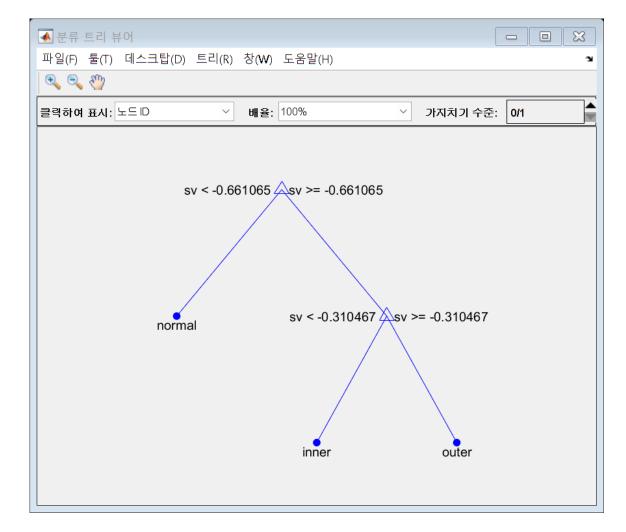




```
pt = prune(t, 'Level', best_level);
cost(best_level+1)

ans = 0
```

view(pt,'Mode','graph')



Compare performance with other classification

Look for a random forest model in Matlab

Compared to Random forest, tree is simple but may not be robust to overfit.

Random forest model is tree-based and it is more robust and stable but may be slow (Boosting)