

# *ASSIGNMENT 7*

## Probabilistic Approaches

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# 1 Classification of iris flowers

In this part, we implement different classification method and parameters to classify the flower iris in the file "iris.arff" .

## 1.1 BFTree

With default parameters, we get the accuracy 94.7%.

Correctly Classified Instances	142	94.6667 %
Incorrectly Classified Instances	8	5.3333 %

We try different parameters to obtain better results. We find when we set the minNumObj as 3, the results become better. The accuracy is 96%.

Correctly Classified Instances	144	96 %
Incorrectly Classified Instances	6	4 %

The classification rules are shown below:

Best-First Decision Tree

```
PL < 2.45: Iris_setosa(50.0/0.0)
PL >= 2.45
|  PW < 1.75
|  |  PL < 4.95: Iris_versicolor(47.0/1.0)
|  |  PL >= 4.95: Iris_virginica(4.0/2.0)
|  PW >= 1.75: Iris_virginica(45.0/1.0)
```

## 1.2 FT

With default parameters, we get the accuracy 96.7%. When we use difference parameters, the result does not become better.

Correctly Classified Instances	145	96.6667 %
Incorrectly Classified Instances	5	3.3333 %

The classification rules is shown below:

=== Classifier model (full training set) ===

FT tree

-----  
: FT\_1:15/15 (150)

Number of Leaves : 1

Size of the Tree : 1

FT\_1:

Class 0 :

23.67 +

[PL] \* -7.42 +

[PW] \* -5.71

Class 1 :

-5.82 +

[SL] \* 1.67 +

[SW] \* 0.11 +

[PL] \* -0.4 +

[PW] \* -1.28

Class 2 :

-31.02 +

[SL] \* -0.4 +

[SW] \* -3.32 +

[PL] \* 5.56 +

[PW] \* 9.9

### 1.3 J48

With default parameters, we get the accuracy 96%. When we use difference parameters, the result does not become better.

Correctly Classified Instances	144	96	%
Incorrectly Classified Instances	6	4	%

The classification rules are shown below:

J48 pruned tree

-----  
PW <= 0.6: Iris\_setosa (50.0)

PW > 0.6

| PW <= 1.7

| | PL <= 4.9: Iris\_versicolor (48.0/1.0)

| | PL > 4.9

| | | PW <= 1.5: Iris\_virginica (3.0)

| | | PW > 1.5: Iris\_versicolor (3.0/1.0)

| PW > 1.7: Iris\_virginica (46.0/1.0)

## 1.4 LADTree

With default parameters, we get the accuracy 94%.

Correctly Classified Instances	141	94	%
Incorrectly Classified Instances	9	6	%

The accuracy can be improved to 95.3% if we change numOfBoostingIterations from 10 to 30.

Correctly Classified Instances	143	95.3333	%
Incorrectly Classified Instances	7	4.6667	%

The classification rules are shown below:

weka.classifiers.trees.LADTree:

```
: 0,0,0
| (1)PW < 0.8: 2,-1,-1
| (1)PW >= 0.8: -1,0.5,0.5
| | (2)PW < 1.75: -0.584,1.39,-0.805
| | | (16)PL < 5.35: -0.445,0.669,-0.224
| | | (16)PL >= 5.35: -0.444,-0.448,0.892
| | (2)PW >= 1.75: -0.584,-0.996,1.58
| | (10)PL < 4.45: -0.443,1.055,-0.612
| | (10)PL >= 4.45: -0.457,-0.185,0.642
| | (13)PL < 4.85: -0.448,0.687,-0.239
| | (13)PL >= 4.85: -0.448,-0.089,0.537
| | (14)PL < 4.95: -0.89,1.287,-0.397
| | (14)PL >= 4.95: -0.89,-0.267,1.158
| | | (15)PW < 1.55: -0.444,-0.449,0.893
| | | (15)PW >= 1.55: -0.445,0.567,-0.122
| | | | (17)PW < 1.75: -0.444,0.774,-0.329
| | | | | (18)SL < 6.95: -0.444,0.89,-0.445
| | | | | (18)SL >= 6.95: -0.444,-0.446,0.89
| | | | (17)PW >= 1.75: -0.444,-0.445,0.889
| (3)PL < 4.95: 0.355,0.256,-0.611
| | (4)PL < 2.45: 0.931,-0.474,-0.457
| | (4)PL >= 2.45: -0.539,0.572,-0.033
| | | (5)SL < 4.95: -0.488,-2.064,2.552
| | | (5)SL >= 4.95: -0.501,0.471,0.03
| | | (6)PW < 1.65: -1.322,3.015,-1.694
| | | (6)PW >= 1.65: -1.352,-0.911,2.263
| | | | (7)SW < 3.1: -1.782,-1.896,3.678
| | | | (7)SW >= 3.1: -1.78,5.609,-3.828
| (3)PL >= 4.95: -0.447,-0.858,1.305
| | (8)PW < 1.55: -0.476,-1.117,1.593
| | (8)PW >= 1.55: -0.497,0.279,0.218
| | | (9)PL < 5.15: -0.917,1.91,-0.992
| | | | (11)PW < 1.75: -1.335,2.925,-1.589
| | | | (11)PW >= 1.75: -1.334,-1.358,2.692
| | | (9)PL >= 5.15: -0.883,-1.232,2.115
```

```
| (12)PL < 2.45: 2.681,-1.343,-1.338
| (12)PL >= 2.45: -1.352,0.406,0.946
| (19)PW < 1.35: 0.875,-0.417,-0.458
| (19)PW >= 1.35: -0.445,-0.076,0.52
```

## 1.5 LMT

With default parameters, we get the accuracy 94%.

Correctly Classified Instances	141	94	%
Incorrectly Classified Instances	9	6	%

Then We change numOfBoostingIterations from -1 to 15 and minNumInstances from 15 to 39, the result becomes better. The accuracy is 98%.

Correctly Classified Instances	147	98	%
Incorrectly Classified Instances	3	2	%

The classification rules are shown below:

Logistic model tree

: LM\_1:15/15 (150)

Number of Leaves : 1

Size of the Tree : 1

LM\_1:

Class 0 :

23.67 +  
[PL] \* -7.42 +  
[PW] \* -5.71

Class 1 :

-5.82 +  
[SL] \* 1.67 +  
[SW] \* 0.11 +  
[PL] \* -0.4 +  
[PW] \* -1.28

Class 2 :

-31.02 +  
[SL] \* -0.4 +  
[SW] \* -3.32 +  
[PL] \* 5.56 +  
[PW] \* 9.9

## 1.6 NBTree

With default parameters, we get the accuracy 94.7%. This method does not have other parameters.

Correctly Classified Instances	142	94.6667	%
Incorrectly Classified Instances	8	5.3333	%

The classification rules are shown below:

NBTree

```
-----
SW <= 3.35
| SW <= 2.95
| | PL <= 4.75: NB 3
| | PL > 4.75: NB 4
| SW > 2.95: NB 5
SW > 3.35
| SL <= 5.9: NB 7
| SL > 5.9: NB 8
```

## 1.7 RandomTree

We get the accuracy 91.3% using default parameters.

Correctly Classified Instances	137	91.3333 %
Incorrectly Classified Instances	13	8.6667 %

Then we change maxDepth from 0 to 10, minNum from 1.0 to 8.0 and seed from 1 to 10, the accuracy is improved to 95.3%.

Correctly Classified Instances	143	95.3333 %
Incorrectly Classified Instances	7	4.6667 %

The classification rules are shown below:

RandomTree  
=====

```
PL < 2.45 : Iris_setosa (50/0)
PL >= 2.45
| PW < 1.75
| | PL < 4.95
| | | PW < 1.65 : Iris_versicolor (47/0)
| | | PW >= 1.65 : Iris_virginica (1/0)
| | PL >= 4.95 : Iris_virginica (6/2)
| PW >= 1.75
| | PL < 4.85 : Iris_virginica (3/1)
| | PL >= 4.85 : Iris_virginica (43/0)
```

## 1.8 REPTree

We get the accuracy 94% using default parameters.

Correctly Classified Instances	141	94	%
Incorrectly Classified Instances	9	6	%

When we try difference parameters, the accuracy does not increase.

The classification rules are shown below:

## REPTree

=====

```
PL < 2.5 : Iris_setosa (33/0) [17/0]
PL >= 2.5
|   PW < 1.75 : Iris_versicolor (36/3) [18/2]
|   PW >= 1.75 : Iris_virginica (31/1) [15/0]
```

## 1.9 SimpleCart

We get the accuracy of 95.3% using default parameters.

Correctly Classified Instances	143	95.3333 %
Incorrectly Classified Instances	7	4.6667 %

Other different parameters cannot make accuracy better.

The classification rules are shown below:

### CART Decision Tree

```
PL < 2.45: Iris_setosa(50.0/0.0)
PL >= 2.45
|   PW < 1.75
|   |   PL < 4.95: Iris_versicolor(47.0/1.0)
|   |   PL >= 4.95
|   |   |   PW < 1.55: Iris_virginica(3.0/0.0)
|   |   |   PW >= 1.55: Iris_versicolor(2.0/1.0)
|   |   PW >= 1.75: Iris_virginica(45.0/1.0)
```

## 1.10 summary

According to all analysis above, it turns out that we get the best accuracy when we implement the LMT method. Hence the highest accuracy we can get is 98%.

# 2 Classification of Congressmen

In this part, we implement different classification method and parameters to classify the congressmen in the file “house-votes-84.arff” .

## 2.1 ADTree

Using default parameters, we can get the accuracy 95.86%.

Correctly Classified Instances	417	95.8621 %
Incorrectly Classified Instances	18	4.1379 %

After we change numOfBoostingIterations from 10 to 25, the result becomes better. The accuracy is improved to 96.32%.

Correctly Classified Instances	419	96.3218 %
Incorrectly Classified Instances	16	3.6782 %

The classification rules are shown below:

Alternating decision tree:

```
: 0.231
| (1)physician_fee_freeze = y: -1.417
| | (5)adoption_of_the_budget_resolution = n: -0.558
| | | (8)superfund_right_to_sue = y: -0.62
| | | | (16)water_project_cost_sharing = n: 0.14
| | | | (16)water_project_cost_sharing != n: -0.688
| | | | | (17)export_administration_act_sa = w: 0.281
| | | | | (17)export_administration_act_sa != w: -0.619
| | | (8)superfund_right_to_sue != y: 0.59
| | (5)adoption_of_the_budget_resolution != n: 0.518
| | | (7)nti_satellite_test_ban = n: 0.984
| | | | (12)water_project_cost_sharing = y: 0.79
| | | | (12)water_project_cost_sharing != y: -0.469
| | | (7)nti_satellite_test_ban != n: -0.981
| | | (19)superfund_right_to_sue = y: 0.359
| | | (19)superfund_right_to_sue != y: -0.518
| | (6)immigration = n: 0.442
| | (6)immigration != n: -0.829
| | (13)synfuels_corporation_cutback = y: 0.156
| | (13)synfuels_corporation_cutback != y: -0.52
| | (20)export_administration_act_sa = y: -0.416
| | (20)export_administration_act_sa != y: 0.19
| (1)physician_fee_freeze != y: 1.66
| | (2)adoption_of_the_budget_resolution = y: 2.381
| | (2)adoption_of_the_budget_resolution != y: -0.956
| (3)synfuels_corporation_cutback = y: 0.914
| | (9)export_administration_act_sa = y: 0.011
| | (9)export_administration_act_sa != y: 0.791
| (3)synfuels_corporation_cutback != y: -0.566
| | (21)water_project_cost_sharing = n: -0.363
| | (21)water_project_cost_sharing != n: 0.161
| (4)education_spending = n: 0.684
| | (11)duty_free_exports = n: -0.314
| | (11)duty_free_exports != n: 0.558
| | (15)immigration = y: -0.139
```



```

| | (15)immigration != y: 0.514
| | | (18)aid_to_nicaraguan_contras = w: -0.229
| | | (18)aid_to_nicaraguan_contras != w: 0.526
| (4)education_spending != n: -0.346
| | (10)physician_fee_freeze = n: 0.614
| | | (14)education_spending = y: 0.682
| | | (14)education_spending != y: -0.392
| | (10)physician_fee_freeze != n: -0.381
| (22)mx_missile = y: 0.344
| | (24)duty_free_exports = n: -0.249
| | (24)duty_free_exports != n: 0.554
| (22)mx_missile != y: -0.119
| | (23)education_spending = w: -0.619
| | (23)education_spending != w: 0.187

```

## 2.2 BFTree

With default parameters, we get the accuracy of 95.17%.

Correctly Classified Instances	414	95.1724 %
Incorrectly Classified Instances	21	4.8276 %

The accuracy increases to 95.86% when we change the minNumObj from 2 to 5 and seed from 1 to 2.

Correctly Classified Instances	417	95.8621 %
Incorrectly Classified Instances	18	4.1379 %

The classification rules are shown below:

Best-First Decision Tree

```

physician_fee_freeze=(n)|(w): democrat(253.0/5.0)
physician_fee_freeze!=(n)|(w): republican(163.0/14.0)

```

## 2.3 DecisionStump

We get the accuracy 95.63 using default parameters. This method does not have other parameters.

Correctly Classified Instances	416	95.6322 %
Incorrectly Classified Instances	19	4.3678 %

The classification rules are shown below:

Decision Stump

Classifications

```
physician_fee_freeze = y : republican
physician_fee_freeze != y : democrat
physician_fee_freeze is missing : democrat
```

Class distributions

```
physician_fee_freeze = y
republican      democrat
0.9209039548022598    0.07909604519774012
physician_fee_freeze != y
republican      democrat
0.01937984496124031    0.9806201550387597
physician_fee_freeze is missing
republican      democrat
0.38620689655172413    0.6137931034482759
```

## 2.4 FT

We get the accuracy 96.78% with default parameters.

Correctly Classified Instances	421	96.7816 %
Incorrectly Classified Instances	14	3.2184 %

The classification rules are shown below:

FT tree

-----

```
N0#1 <= 0.595535: FT_1:15/30 (268)
N0#1 > 0.595535: FT_2:15/30 (167)
```

## 2.5 Id3

The accuracy is 94.25% with default parameters.

Correctly Classified Instances	410	94.2529 %
Incorrectly Classified Instances	22	5.0575 %

The classification rules are shown below:

```

physician_fee_freeze = y
| synfuels_corporation_cutback = w: republican
| synfuels_corporation_cutback = n
| | duty_free_exports = n
| | | adoption_of_the_budget_resolution = n: republican
| | | adoption_of_the_budget_resolution = y
| | | | export_administration_act_sa = y: republican
| | | | export_administration_act_sa = w
| | | | | handicapped_infants = n: democrat
| | | | | handicapped_infants = w: null
| | | | | handicapped_infants = y: republican
| | | | export_administration_act_sa = n: null
| | | adoption_of_the_budget_resolution = w: null
| | duty_free_exports = y
| | | immigration = y: republican
| | | immigration = n
| | | | export_administration_act_sa = y: democrat
| | | | export_administration_act_sa = w
| | | | | water_project_cost_sharing = y: republican
| | | | | water_project_cost_sharing = n: democrat
| | | | | water_project_cost_sharing = w: null
| | | | export_administration_act_sa = n: republican
| | | immigration = w: null
| | duty_free_exports = w: republican
| synfuels_corporation_cutback = y

| | adoption_of_the_budget_resolution = n
| | | el_salvador_aid = y
| | | | export_administration_act_sa = y: republican
| | | | export_administration_act_sa = w
| | | | | handicapped_infants = n: republican
| | | | | handicapped_infants = w: null
| | | | | handicapped_infants = y: democrat
| | | | export_administration_act_sa = n
| | | | | superfund_right_to_sue = y
| | | | | | water_project_cost_sharing = y: republican
| | | | | | water_project_cost_sharing = n
| | | | | | | handicapped_infants = n: democrat
| | | | | | | handicapped_infants = w: null
| | | | | | | handicapped_infants = y: republican
| | | | | | water_project_cost_sharing = w: null
| | | | | superfund_right_to_sue = w: null
| | | | | superfund_right_to_sue = n: democrat
| | | | el_salvador_aid = w: null
| | | | el_salvador_aid = n: democrat
| | | adoption_of_the_budget_resolution = y
| | | | nti_satellite_test_ban = n: democrat
| | | | nti_satellite_test_ban = y: republican
| | | | nti_satellite_test_ban = w: null
| | | adoption_of_the_budget_resolution = w: democrat
physician_fee_freeze = w
| mx_missile = n: democrat
| mx_missile = y
| | nti_satellite_test_ban = n: republican
| | nti_satellite_test_ban = y: democrat
| | nti_satellite_test_ban = w: democrat
| mx_missile = w: republican

```

```

physician_fee_freeze = n
| adoption_of_the_budget_resolution = n
| | education_spending = y: democrat
| | education_spending = n
| | | synfuels_corporation_cutback = w: null
| | | synfuels_corporation_cutback = n
| | | | religious_groups_in_schools = y: democrat
| | | | religious_groups_in_schools = n
| | | | crime = y: republican
| | | | crime = n: democrat
| | | | crime = w: null
| | | | religious_groups_in_schools = w: null
| | | synfuels_corporation_cutback = y: democrat
| | education_spending = w: republican
| adoption_of_the_budget_resolution = y: democrat
| adoption_of_the_budget_resolution = w: democrat

```

## 2.6 J48

The accuracy is 94.94% with default parameters.

Correctly Classified Instances	413	94.9425 %
Incorrectly Classified Instances	22	5.0575 %

The accuracy can be improved to 96.32% when we change the minNumObj from 2 to 7 and confidenceFactor from 0.25 to 1.

Correctly Classified Instances	419	96.3218 %
Incorrectly Classified Instances	16	3.6782 %

The classification rules are shown below:

J48 pruned tree

-----

```

physician_fee_freeze = y
| synfuels_corporation_cutback = w: republican (7.0)
| synfuels_corporation_cutback = n: republican (138.0/3.0)
| synfuels_corporation_cutback = y
| | adoption_of_the_budget_resolution = n: republican (23.0/5.0)
| | adoption_of_the_budget_resolution = y: democrat (8.0/3.0)
| | adoption_of_the_budget_resolution = w: democrat (1.0)
physician_fee_freeze = w: democrat (11.0/3.0)
physician_fee_freeze = n
| adoption_of_the_budget_resolution = n
| | education_spending = y: democrat (10.0)
| | education_spending = n: democrat (14.0/1.0)
| | education_spending = w: republican (1.0)
| adoption_of_the_budget_resolution = y: democrat (219.0)
| adoption_of_the_budget_resolution = w: democrat (3.0)

```

## 2.7 LADTree

The accuracy is 94.94% with default parameters.

Correctly Classified Instances	413	94.9425 %
Incorrectly Classified Instances	22	5.0575 %

The accuracy can be increased to 96.32% when we change the numOfBoostingIterations from 10 to 20.

Correctly Classified Instances	419	96.3218 %
Incorrectly Classified Instances	16	3.6782 %

The classification rules is shown below(part):

```
: 0,0
| (1)physician_fee_freeze = y: 0.977,-0.977
| | (6)immigration = n: -0.34,0.34
| | | (9)mx_missile = n: 0.201,-0.201
| | | | (10)superfund_right_to_sue = y: 0.236,-0.236
| | | | | (13)export_administration_act_sa = y: 0.571,-0.571
| | | | | (13)export_administration_act_sa != y: -0.284,0.284
| | | | (10)superfund_right_to_sue != y: -0.987,0.987
| | | | | (15)export_administration_act_sa = w: 0.588,-0.588
| | | | | (15)export_administration_act_sa != w: -0.613,0.613
| | | (9)mx_missile != n: -0.847,0.847
| | (6)immigration != n: 0.532,-0.532
| (1)physician_fee_freeze != y: -1.236,1.236
| (2)adoption_of_the_budget_resolution = y: -0.529,0.529
| | (7)nti_satellite_test_ban = y: 0.071,-0.071
| | | (8)physician_fee_freeze = y: 1.403,-1.403
| | | (8)physician_fee_freeze != y: -1.016,1.016
| | (7)nti_satellite_test_ban != y: -0.878,0.878
| | | (11)water_project_cost_sharing = y: -0.679,0.679
| | | (11)water_project_cost_sharing != y: 0.632,-0.632
| (2)adoption_of_the_budget_resolution != y: 0.345,-0.345
| (3)synfuels_corporation_cutback = y: -0.61,0.61
| (3)synfuels_corporation_cutback != y: 0.217,-0.217
| | (4)physician_fee_freeze = n: -0.491,0.491
| | | (12)superfund_right_to_sue = n: 0.573,-0.573
| | | | (16)adoption_of_the_budget_resolution = n: 0.716,-0.716
| | | | (16)adoption_of_the_budget_resolution != n: -0.512,0.512
| | | (12)superfund_right_to_sue != n: -0.605,0.605
| | (4)physician_fee_freeze != n: 0.452,-0.452
| | (14)education_spending = n: -0.183,0.183
| | | (17)religious_groups_in_schools = n: 0.465,-0.465
| | | | (18)crime = y: 0.837,-0.837
| | | | (18)crime != y: -0.786,0.786
| | | (17)religious_groups_in_schools != n: -0.376,0.376
| | (14)education_spending != n: 0.352,-0.352
| (5)education_spending = n: -0.435,0.435
| (5)education_spending != n: 0.191,-0.191
```

## 2.8 LMT

We get the accuracy 96.78% using default parameters,

Correctly Classified Instances	421	96.7816 %
Incorrectly Classified Instances	14	3.2184 %

It makes no difference when we use different parameters.

The classification rules are shown below:

```
Logistic model tree
-----
: LM_1:20/20 (435)

Number of Leaves :    1

Size of the Tree :    1
LM_1:
Class 0 :
0.31 +
[water_project_cost_sharing=y] * -0.51 +
[adoption_of_the_budget_resolution=n] * 0.38 +
[adoption_of_the_budget_resolution=y] * -0.85 +
[physician_fee_freeze=y] * 1.8 +
[physician_fee_freeze=n] * -1.58 +
[nti_satellite_test_ban=n] * -0.38 +
[nti_satellite_test_ban=y] * 0.32 +
[aid_to_nicaraguan_contras=w] * 0.44 +
[mx_missile=y] * -0.54 +
[immigration=n] * -0.59 +
[synfuels_corporation_cutback=n] * 0.71 +
[synfuels_corporation_cutback=y] * -0.8 +
[education_spending=n] * -0.58 +
[education_spending=w] * 0.35 +
[superfund_right_to_sue=w] * -0.46 +
[duty_free_exports=y] * -0.41 +
[export_administration_act_sa=y] * 0.41

Class 1 :
-0.31 +
[water_project_cost_sharing=y] * 0.51 +
[adoption_of_the_budget_resolution=n] * -0.38 +
[adoption_of_the_budget_resolution=y] * 0.85 +
[physician_fee_freeze=y] * -1.8 +
[physician_fee_freeze=n] * 1.58 +
[nti_satellite_test_ban=n] * 0.38 +
[nti_satellite_test_ban=y] * -0.32 +
[aid_to_nicaraguan_contras=w] * -0.44 +
[mx_missile=y] * 0.54 +
[immigration=n] * 0.59 +
[synfuels_corporation_cutback=n] * -0.71 +
[synfuels_corporation_cutback=y] * 0.8 +
[education_spending=n] * 0.58 +
[education_spending=w] * -0.35 +
[superfund_right_to_sue=w] * 0.46 +
[duty_free_exports=y] * 0.41 +
[export_administration_act_sa=y] * -0.41
```

## 2.9 NBTree

The accuracy is 94.71% when we use default parameters.

Correctly Classified Instances	412	94.7126 %
Incorrectly Classified Instances	23	5.2874 %

The classification rules are shown below:

NBTree

-----

```
nti_satellite_test_ban = n
|   aid_to_nicaraguan_contras = n: NB 2
|   aid_to_nicaraguan_contras = y: NB 3
|   aid_to_nicaraguan_contras = w: NB 4
nti_satellite_test_ban = y
|   superfund_right_to_sue = y: NB 6
|   superfund_right_to_sue = w: NB 7
|   superfund_right_to_sue = n
|   |   physician_fee_freeze = y: NB 9
|   |   physician_fee_freeze = w: NB 10
|   |   physician_fee_freeze = n
|   |   |   synfuels_corporation_cutback = w: NB 12
|   |   |   synfuels_corporation_cutback = n: NB 13
|   |   |   synfuels_corporation_cutback = y: NB 14
nti_satellite_test_ban = w: NB 15
```

## 2.10 RandomTree

The accuracy is 93.56% with default parameters.

Correctly Classified Instances	407	93.5632 %
Incorrectly Classified Instances	28	6.4368 %

The accuracy can be increased to 94.71% when we change the seed from 1 to 15.

Correctly Classified Instances	410	94.2529 %
Incorrectly Classified Instances	25	5.7471 %

The classification rules are shown below(part):

## RandomTree

=====

```

aid_to_nicaraguan_contras = n
|   education_spending = y
|   |   synfuels_corporation_cutback = w : republican (3/0)
|   |   synfuels_corporation_cutback = n
|   |   |   physician_fee_freeze = y
|   |   |   |   adoption_of_the_budget_resolution = n : republican (90/0)
|   |   |   |   adoption_of_the_budget_resolution = y
|   |   |   |   |   water_project_cost_sharing = y
|   |   |   |   |   |   export_administration_act_sa = y : republican (3/0)
|   |   |   |   |   |   export_administration_act_sa = w : democrat (1/0)
|   |   |   |   |   |   export_administration_act_sa = n : republican (0/0)
|   |   |   |   |   water_project_cost_sharing = n : republican (3/0)
|   |   |   |   |   water_project_cost_sharing = w : republican (0/0)
|   |   |   |   |   adoption_of_the_budget_resolution = w : republican (1/0)
|   |   |   |   physician_fee_freeze = w : republican (0/0)
|   |   |   |   physician_fee_freeze = n : democrat (3/0)
|   |   |   synfuels_corporation_cutback = y
|   |   |   |   export_administration_act_sa = y
|   |   |   |   |   adoption_of_the_budget_resolution = n
|   |   |   |   |   |   superfund_right_to_sue = y
|   |   |   |   |   |   |   immigration = y
|   |   |   |   |   |   |   |   water_project_cost_sharing = y : republican (3/0)
|   |   |   |   |   |   |   |   water_project_cost_sharing = n : democrat (1/0)
|   |   |   |   |   |   |   |   water_project_cost_sharing = w : republican (0/0)
|   |   |   |   |   |   |   immigration = n : republican (4/0)
|   |   |   |   |   |   |   immigration = w : republican (0/0)
|   |   |   |   |   |   superfund_right_to_sue = w : republican (0/0)
|   |   |   |   |   |   superfund_right_to_sue = n
|   |   |   |   |   |   |   handicapped_infants = n : republican (1/0)
|   |   |   |   |   |   |   handicapped_infants = w : republican (0/0)
|   |   |   |   |   |   |   handicapped_infants = y : democrat (1/0)

```

## 2.11 SimpleCart

The accuracy is 95.63% with default parameters. This accuracy is not improved under different parameters.

Correctly Classified Instances	416	95.6322 %
Incorrectly Classified Instances	19	4.3678 %

The classification rules are shown below:



## CART Decision Tree

```
physician_fee_freeze=(n)|(w)
| adoption_of_the_budget_resolution=(y)|(n): democrat(247.0/2.0)
| adoption_of_the_budget_resolution!=(y)|(n)
| | mx_missile=(n)|(y): democrat(6.0/1.0)
| | mx_missile!=(n)|(y): republican(2.0/0.0)
physician_fee_freeze!=(n)|(w)
| synfuels_corporation_cutback=(y)
| | adoption_of_the_budget_resolution=(w)|(y)
| | | nti_satellite_test_ban=(n)|(w): democrat(6.0/0.0)
| | | nti_satellite_test_ban!=(n)|(w): republican(3.0/0.0)
| | adoption_of_the_budget_resolution!=(w)|(y)
| | | el_salvador_aid=(n): democrat(2.0/0.0)
| | | el_salvador_aid!=(n): republican(18.0/3.0)
| synfuels_corporation_cutback!=(y): republican(142.0/3.0)
```

## 2.12 summary

According to all analysis above, it turns out that we get the best accuracy when we implement LMT and FT methods. Hence the highest accuracy is 96.78%

## 3 Observation and Conclusion

As it can be seen, LMT is the best method for the classification of Iris among all kinds of methods. As for the classification of Congressmen, LMT and FT are best methods. We obtain best accuracy in both classification using LMT method. In this way, LMT method has the best performance in this project.