

# Supervised Learning on Bakery Data Using WEKA

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CSE 352

# Outline

- Classification Tool: **WEKA**
- **Waikato Environment for Knowledge Analysis** by The University of Waikato.
- Available on the internet at:  
<http://www.cs.waikato.ac.nz/~ml/weka/index.html>





# Raw Data

- The Raw data does give us a lot of information.
- However, in this form most of this information is useless and doesn't tell us anything.

K	L	M	N	O	P	Q	R	S
0.80	0.02	4318	7		4		0.1	7.5
0.48	0.04	29649	10	26	201	62.3	0.7	10.7
0.01	0.00	5726	14	2	7005		0.4	4.1
0.40	0.01	29379	13		6407		1.7	10.1
0.03	0.03	2158	1919	6112	3	9.2	7.9	
0.07	0.00	324570	13715	10617	42	39.4	57.1	9.6
0.01	0.02	16244	31969	106331	302	116.2	112.5	
0.01	0.06	1632	2314	6	6		12.5	2.7
0.09	0.00	791	2477	6	2		5.8	4.5
0.00	0.01	13165	12521	5277	22	59.5	47.0	
0.84	0.06	3321	25	7	9	9.11	1.6	20.1
2.45	0.06	2888	111	21	51	80.5	0.6	115.5
4.23	0.01	209851	23	76	331	121	3.0	79.7
4.98	0.06	38791	73	27	96	22.1	1.2	64.7
8.03	0.08	13769	11	11	5	25.5	0.5	105.2
2.65	0.08	275	37		4		0.6	49.9
3.45	0.03	184	13	7	2		0.6	6.8
9.65	0.08	652	10	6	288	185	0.7	60.1
1.37	0.03	314	9		5			16.6
0.66	0.09	1222	16	3	5		0.3	13.1
K2O	P2O5	S	Zn	Pb	Cu	As	Cd	Cr

# Data Preparation

- To prepare data for pre-processing the following steps were taken.
- Any attributes that have missing data (i.e. more than 20%) will be removed.
- The following attributes were thus removed:
  - Pb
  - As
  - Cd
  - Ni
  - Sc
  - Co
  - Li
  - Mo



# Data Preparation

- All other attributes that are missing values were filled in with their **averages** (mean).
- Missing values for the following attributes were inserted:
  - TiO<sub>2</sub> (Carbonates) – Mean: 0.005 (Inserted at E58 and E62)
  - P<sub>2</sub>O<sub>5</sub> (Carbonates) – Mean: 0.74 (Inserted at L33)
  - S (Carbonates) – Mean: 423 (Inserted at M52, M66, M75)
  - Zn (Carbonates) – Mean: 16 (Inserted at N46, N53, N67)
  - Cu (Carbonates) – Mean: 3 (Inserted at O37, O43, O46, O48, O52, O55, O57, O60, O63, O67)
  - Cr (Galene) – Mean: 9.6 (Inserted at P85, P87)
  - Cr (Spahlerite) – Mean: 3.6 (Inserted at P90)
  - V (Carbonates) – Mean: 5.2 (Inserted at Q43, Q48, Q51, Q52, Q60, Q62, Q66, Q71, Q75)
  - V (Galene) – Mean: 2.5 (Inserted at Q86)
  - V (Spahlerite) – Mean: 9.4 (Inserted at Q90)

# Data Preparation

- For easier reading, class values were replaced with simpler values.
- The following values were changed:
  - R. carbonatées changed to C<sub>1</sub>
  - Pyrite changed to C<sub>2</sub>
  - Chalcopyrites changed to C<sub>3</sub>
  - Galène changed to C<sub>4</sub>
  - Spahlerite changed to C<sub>5</sub>
  - Sédiments terrigènes changed to C<sub>6</sub>



# Data Preparation

- Using WEKA, we remove any **noisy data** that may unnecessarily skew our data and results.

The screenshot shows the WEKA Explorer interface. The 'Filter' section is set to 'RemoveUseless -M 99.0'. The 'Current relation' is 'bakarydata' with 103 instances. The 'Attributes' list includes: 1. Echantillon, 2. Type de roche, 3. CaO+MgO (selected), 4. Al2O3, 5. TiO2, 6. Fe2O3\*, 7. MnO, 8. MgO, 9. CaO, 10. Na2O, 11. K2O, 12. P2O5, 13. K. A histogram at the bottom right shows the distribution of the selected attribute 'CaO+MgO' with values 0.48, 29.51, and 58.55. The histogram bars are colored green, blue, and red, with counts of 7, 1, 2, 7, 51, and 30. The 'About' dialog box for 'weka.filters.unsupervised.attribute.RemoveUseless' is open, displaying the description: 'Removes constant attributes, along with nominal attributes that vary too much.' and the 'maximumVariancePercentageAllowed' set to 99.0.

Weka Explorer

Preprocess | Classify | Cluster | Associate | Select attributes | Visualize

Open file... | Open URL... | Open DB... | Undo | Edit... | Save...

Filter

Choose **RemoveUseless -M 99.0** Apply

Current relation Selected attribute

Relation: bakarydata Name: CaO+MgO Type: Numeric

Instances: 103 Attribute: CaO+MgO (87%)

Attributes

All None

No.	Name
1	Echantillon
2	Type de roche
3	CaO+MgO
4	Al2O3
5	TiO2
6	Fe2O3*
7	MnO
8	MgO
9	CaO
10	Na2O
11	K2O
12	P2O5
13	K

Remove

Visualize All

Status OK Log x 0

weka.gui.GenericObjectEditor

weka.filters.unsupervised.attribute.RemoveUseless

About

Removes constant attributes, along with nominal attributes that vary too much. More

maximumVariancePercentageAllowed 99.0

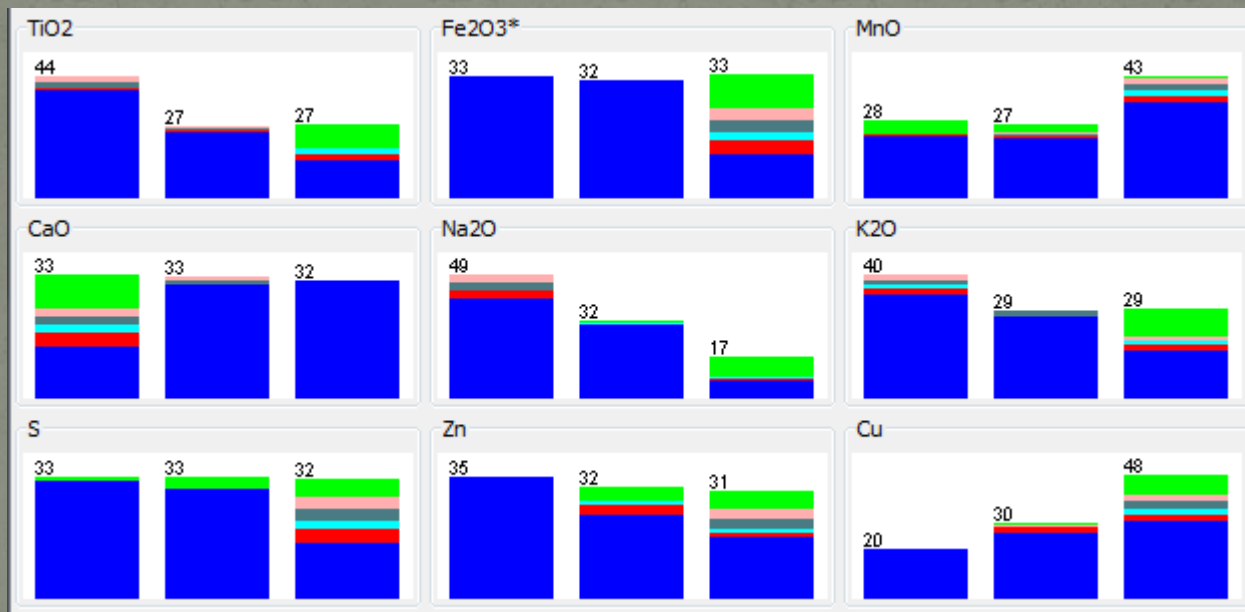
Open... Save... OK Cancel

7 1 2 7 51 30

0.48 29.51 58.55

# Discretization

- With all the missing data filled in, the noisy data eliminated, we discretize the data using the WEKA tool. (3 equal frequency bins)





# Discretization

- Values in the bins were then replaced by specific words:

- Low
- Medium
- High

Label	Count
Low	32
Medium	38
High	28

- This helps in understanding data better.
- Decision Tree algorithms will still work with these non-numerical values.

# Experiments

- The following experiments will be carried out on our data:
  - **Full Learning:** Construction of decision trees characterizing all classes.
  - **Contrast Learning:** Using all attributes to compare class  $C_1$  with the rest of the classes.
  - **Limited Learning:** Construction of decision tree using only the major attributes.

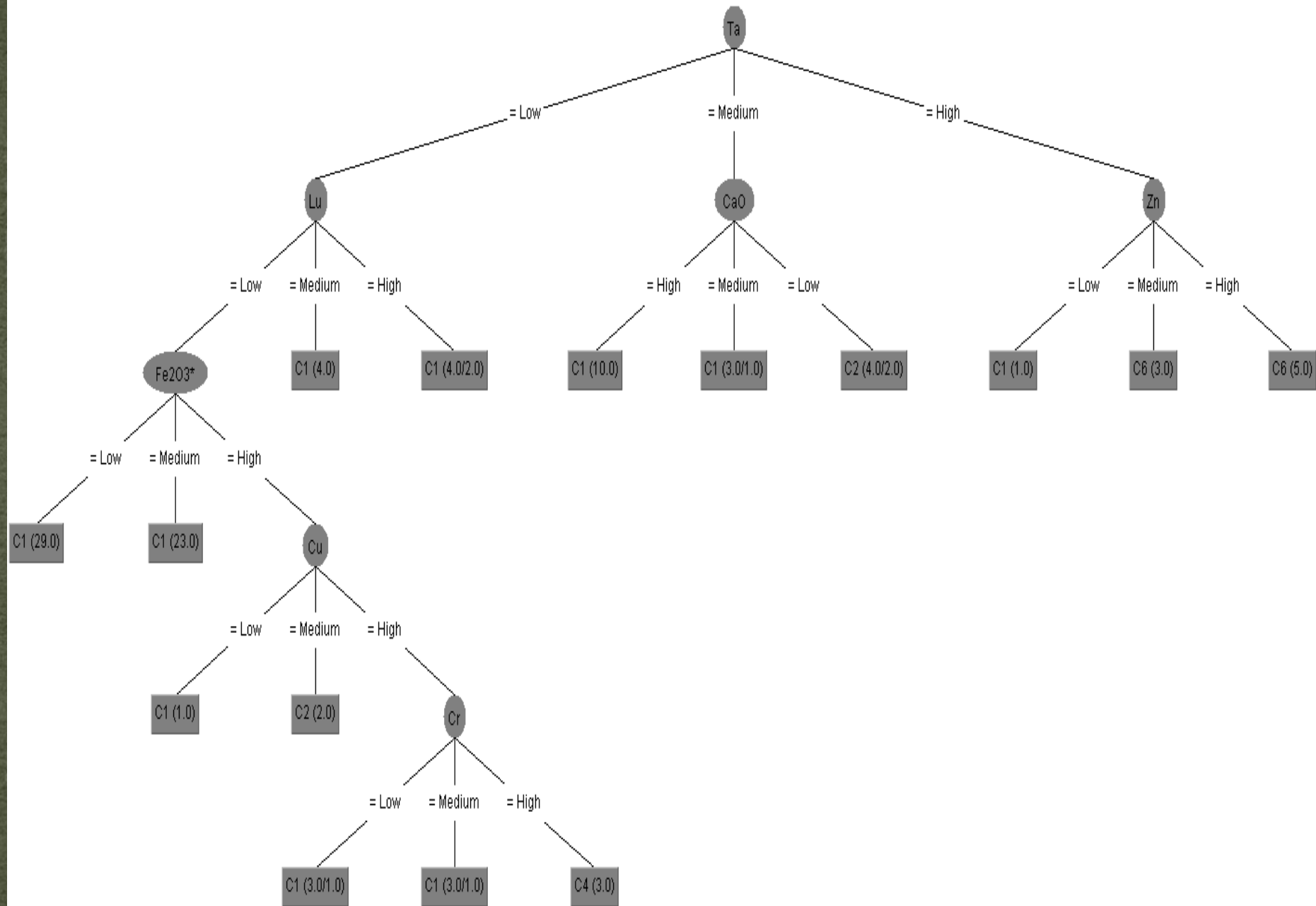


# Experiment 1 - Results

- Experiment 1: Full Learning
- Decision Trees were generated using the J48 algorithm.

```
Classifier output
J48 unpruned tree
-----
Ta = Low
| Lu = Low
| | Fe2O3* = Low: C1 (29.0)
| | Fe2O3* = Medium: C1 (23.0)
| | Fe2O3* = High
| | | Cu = Low: C1 (1.0)
| | | Cu = Medium: C2 (2.0)
| | | Cu = High
| | | | Cr = Low: C1 (3.0/1.0)
| | | | Cr = Medium: C1 (3.0/1.0)
| | | | Cr = High: C4 (3.0)
| Lu = Medium: C1 (4.0)
| Lu = High: C1 (4.0/2.0)
Ta = Medium
| CaO = High: C1 (10.0)
| CaO = Medium: C1 (3.0/1.0)
| CaO = Low: C2 (4.0/2.0)
Ta = High
| Zn = Low: C1 (1.0)
| Zn = Medium: C6 (3.0)
| Zn = High: C6 (5.0)

Number of Leaves :    15
Size of the tree :    22
```





# Discriminant Rules

- We got the following discriminant rules.
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="Low" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="Medium" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="Low" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="Medium" THEN Class="C<sub>2</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="High" AND Cr="Low" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="High" AND Cr="Medium" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="High" AND Cr="High" THEN Class="C<sub>4</sub>"

# More Discriminant Rules

- We got the following discriminant rules.
  - IF Ta="Low" AND Lu="Medium" THEN Class="C1"
  - IF Ta="Low" AND Lu="High" THEN Class="C1"
  - IF Ta="Medium" AND CaO="High" THEN Class="C1"
  - IF Ta="Medium" AND CaO="Medium" THEN Class="C1"
  - IF Ta="Medium" AND CaO="Low" THEN Class="C2"
  - IF Ta="High" AND Zn="Low" THEN Class="C1"
  - IF Ta="High" AND Zn="Medium" THEN Class="C6"
  - IF Ta="High" AND Zn="High" THEN Class="C6"
- Predictive Accuracy Determined: 70.58%



# Experiment 2 - Results

- Experiment 2: Contrast Learning
- Decision Trees were generated using the J48 algorithm.

```
Classifier output
J48 unpruned tree
-----

Fe2O3* = Low: C1 (34.0)
Fe2O3* = Medium: C1 (32.0)
Fe2O3* = High
|   CaO = High: C1 (4.0)
|   CaO = Medium
|   |   Rb = Medium: NOT C1 (3.0/1.0)
|   |   Rb = Low: C1 (1.0)
|   |   Rb = High: C1 (3.0)
|   CaO = Low
|   |   Zn = Low: C1 (2.0)
|   |   Zn = Medium: NOT C1 (8.0)
|   |   Zn = High: NOT C1 (11.0)

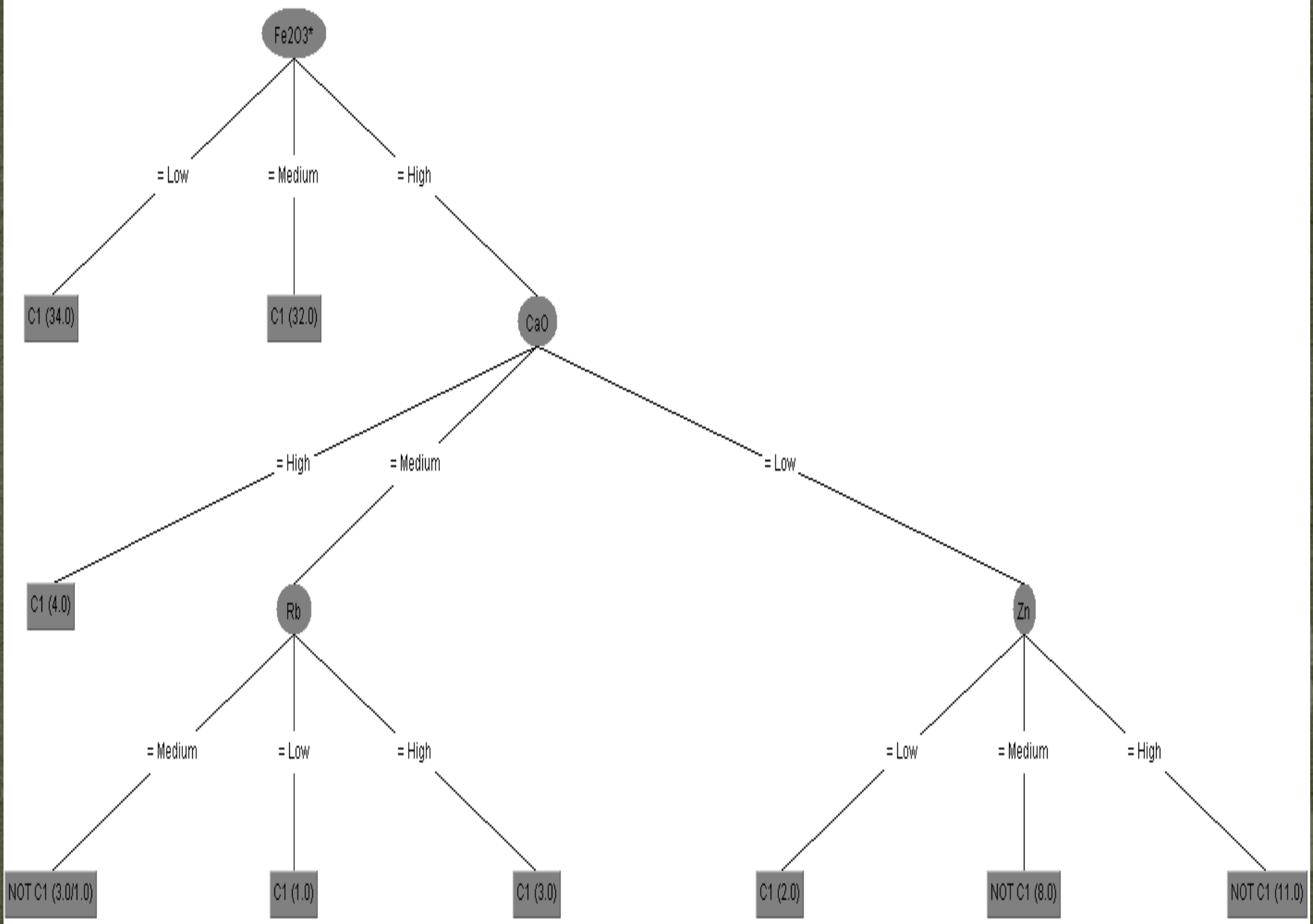
Number of Leaves :      9

Size of the tree :      13

Time taken to build model: 0 seconds

=== Evaluation on test split ===
=== Summary ===

Correctly Classified Instances      33          97.0588 %
Incorrectly Classified Instances     1          2.9412 %
```





# Discriminant Rules

- We got the following discriminant rules.
  - IF  $\text{Fe}_2\text{O}_3$ ="Low" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="Medium" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="High" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Medium" AND Rb="Medium" THEN Class="NOT C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Medium" AND Rb="Low" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Medium" AND Rb="High" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Low" AND Zn="Low" THEN Class="C<sub>1</sub>"

# More Discriminant Rules

- We got the following discriminant rules.
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND  $\text{CaO}$ ="Low" AND  $\text{Zn}$ ="Medium"  
THEN Class="NOT  $C_1$ "
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND  $\text{CaO}$ ="Low" AND  $\text{Zn}$ ="High"  
THEN Class="NOT  $C_1$ "
- Predictive Accuracy Determined: 97.06%

# Experiment 3 - Results

- Experiment 3: Using Major Attributes
- Decision Trees were generated using the J48 algorithm.

```
Classifier output
J48 unpruned tree
-----

Fe2O3* = Low: C1 (34.0)
Fe2O3* = Medium: C1 (32.0)
Fe2O3* = High
| CaO = High: C1 (4.0)
| CaO = Medium: C1 (7.0/2.0)
| CaO = Low
| | Zn = Low: C1 (2.0)
| | Zn = Medium
| | | S = Low: C6 (1.0)
| | | S = Medium: C6 (2.0)
| | | S = High: C2 (5.0/2.0)
| | Zn = High: C6 (11.0/6.0)

Number of Leaves :      9

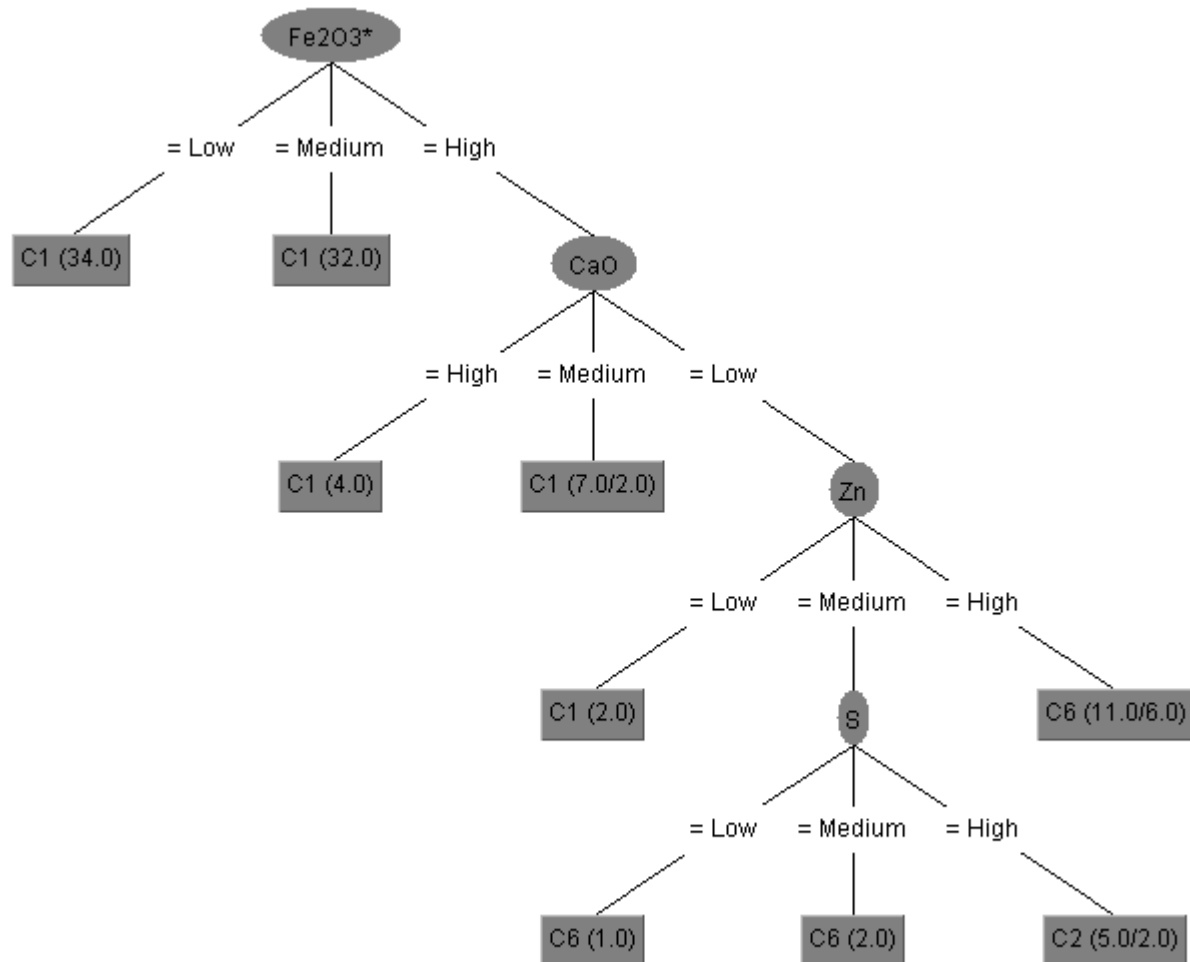
Size of the tree :      13

Time taken to build model: 0 seconds

=== Evaluation on test split ===
=== Summary ===

Correctly Classified Instances      28      82.3529 %
Incorrectly Classified Instances     6      17.6471 %
Kappa statistic                     0.5768
```





# Discriminant Rules

- We got the following discriminant rules.
  - IF  $\text{Fe}_2\text{O}_3$ ="Low" THEN Class="C1"
  - IF  $\text{Fe}_2\text{O}_3$ ="Medium" THEN Class="C1"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="High" THEN Class="C1"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Medium" THEN Class="C1"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Low" AND Zn="Low" THEN Class="C1"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Low" AND Zn="Medium" AND S="Low" THEN Class="C6"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Low" AND Zn="Medium" AND S="Medium" THEN Class="C6"

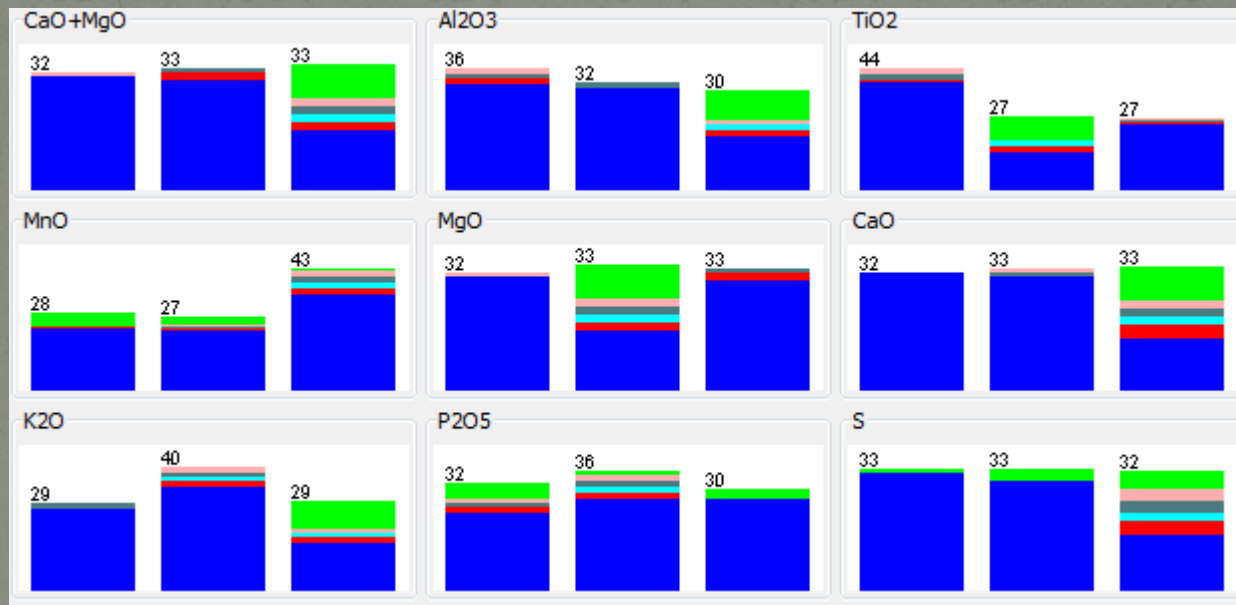
# More Discriminant Rules

- We got the following discriminant rules.
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND  $\text{CaO}$ ="Low" AND  $\text{Zn}$ ="Medium" AND  $\text{S}$ ="High" THEN Class="C2"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND  $\text{CaO}$ ="Low" AND  $\text{Zn}$ ="High" THEN Class="C6"
- Predictive Accuracy Determined: 82.35%



# Discretization for Dataset 2

- With all the missing data filled in, the noisy data eliminated, we use another method of data discretization. (4 equal width bins)



# Experiments with Dataset 2

- The following experiments will be carried out on our data:
  - **Full Learning:** Construction of decision trees characterizing all classes.
  - **Contrast Learning:** Using all attributes to compare class  $C_1$  with the rest of the classes.
  - **Limited Learning:** Construction of decision tree using only the major attributes.

# Experiment 1 Dataset 2- Results

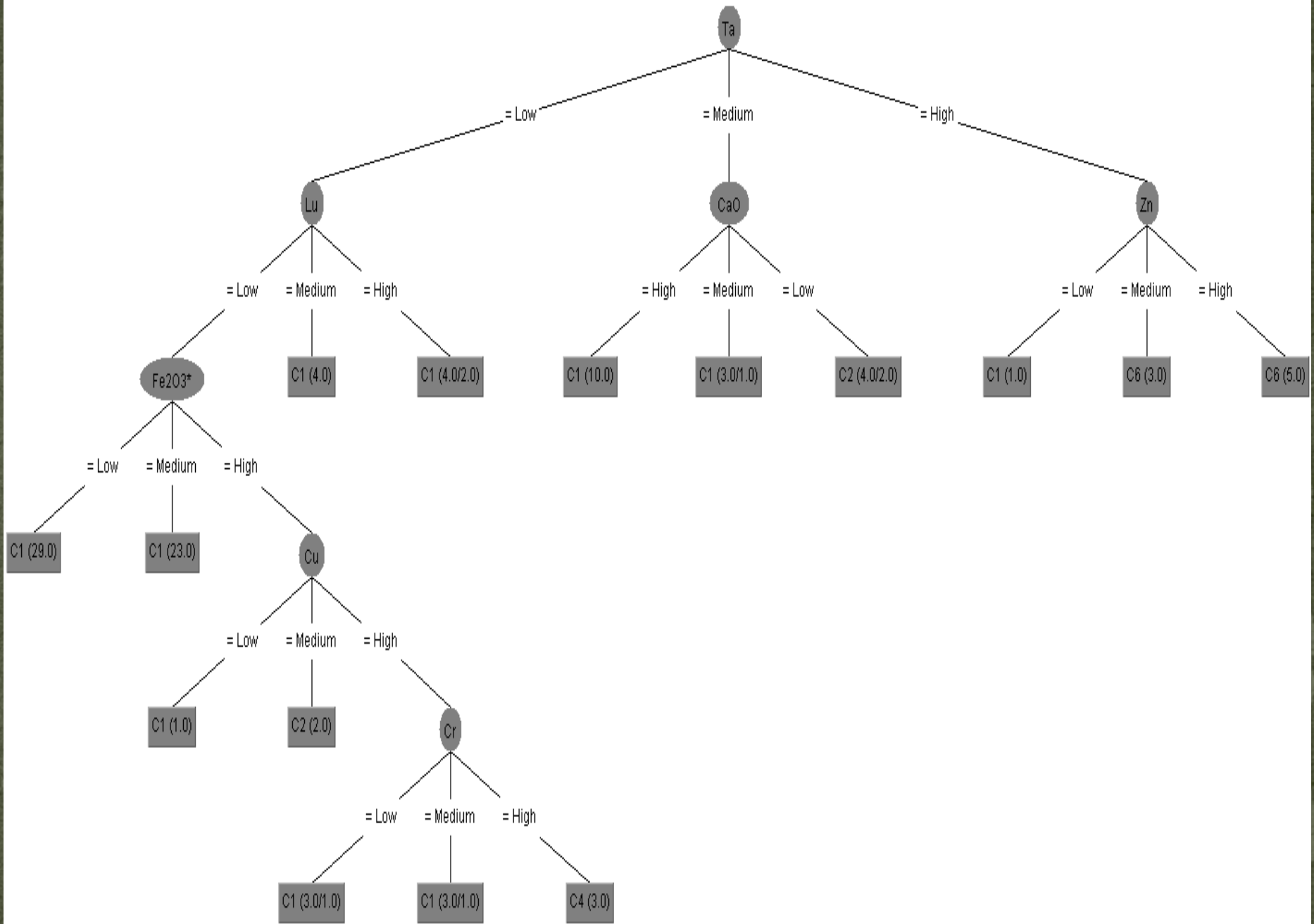
- Experiment 1: Full Learning
- Decision Trees were generated using the J48 algorithm.

```
Classifier output
J48 unpruned tree
-----

Ta = Low
|  Lu = Low
|  |  Fe2O3* = Low: C1 (29.0)
|  |  Fe2O3* = Medium: C1 (23.0)
|  |  Fe2O3* = High
|  |  |  Cu = Low: C1 (1.0)
|  |  |  Cu = Medium: C2 (2.0)
|  |  |  Cu = High
|  |  |  |  Cr = Low: C1 (3.0/1.0)
|  |  |  |  Cr = Medium: C1 (3.0/1.0)
|  |  |  |  Cr = High: C4 (3.0)
|  Lu = Medium: C1 (4.0)
|  Lu = High: C1 (4.0/2.0)
Ta = Medium
|  CaO = High: C1 (10.0)
|  CaO = Medium: C1 (3.0/1.0)
|  CaO = Low: C2 (4.0/2.0)
Ta = High
|  Zn = Low: C1 (1.0)
|  Zn = Medium: C6 (3.0)
|  Zn = High: C6 (5.0)

Number of Leaves :    15
Size of the tree :    22
```





# Discriminant Rules

- We got the following discriminant rules.
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="Low" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="Medium" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="Low" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="Medium" THEN Class="C<sub>2</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="High" AND Cr="Low" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="High" AND Cr="Medium" THEN Class="C<sub>1</sub>"
  - IF Ta="Low" AND Lu="Low" AND Fe<sub>2</sub>O<sub>3</sub>="High" AND Cu="High" AND Cr="High" THEN Class="C<sub>4</sub>"

# More Discriminant Rules

- We got the following discriminant rules.
  - IF Ta="Low" AND Lu="Medium" THEN Class="C1"
  - IF Ta="Low" AND Lu="High" THEN Class="C1"
  - IF Ta="Medium" AND CaO="High" THEN Class="C1"
  - IF Ta="Medium" AND CaO="Medium" THEN Class="C1"
  - IF Ta="Medium" AND CaO="Low" THEN Class="C2"
  - IF Ta="High" AND Zn="Low" THEN Class="C1"
  - IF Ta="High" AND Zn="Medium" THEN Class="C6"
  - IF Ta="High" AND Zn="High" THEN Class="C6"
- Predictive Accuracy Determined: 79.59%



# Experiment 2 Dataset 2 - Results

- Experiment 2: Contrast Learning
- Decision Trees were generated using the J48 algorithm.

```
Classifier output

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

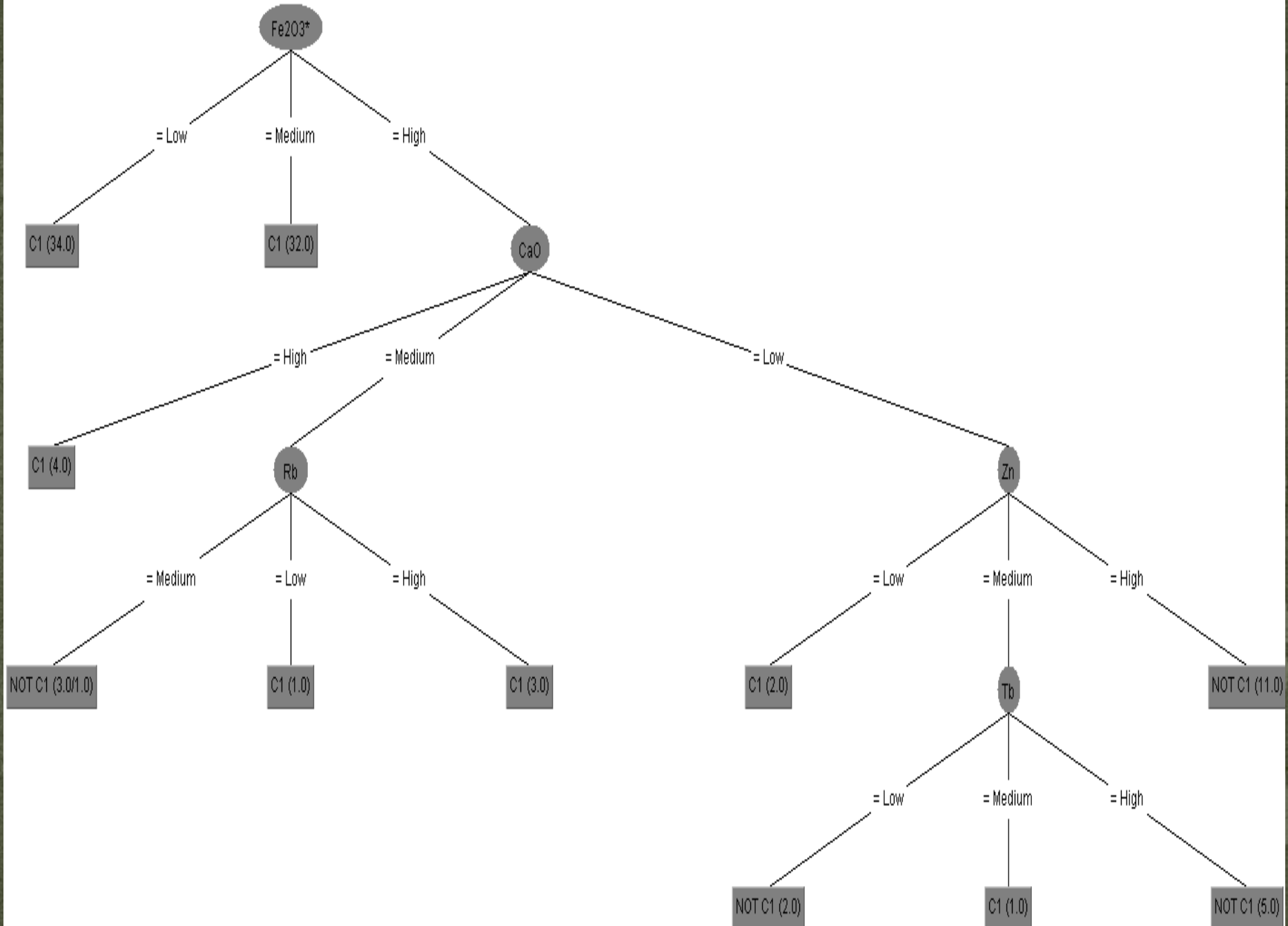
J48 unpruned tree
-----

Fe2O3* = Low: C1 (34.0)
Fe2O3* = Medium: C1 (32.0)
Fe2O3* = High
|   CaO = High: C1 (4.0)
|   CaO = Medium
|   |   Rb = Medium: NOT C1 (3.0/1.0)
|   |   Rb = Low: C1 (1.0)
|   |   Rb = High: C1 (3.0)
|   CaO = Low
|   |   Zn = Low: C1 (2.0)
|   |   Zn = Medium
|   |   |   Tb = Low: NOT C1 (2.0)
|   |   |   Tb = Medium: C1 (1.0)
|   |   |   Tb = High: NOT C1 (5.0)
|   |   Zn = High: NOT C1 (11.0)

Number of Leaves :      11

Size of the tree :      16

Time taken to build model: 0 seconds
```



# Discriminant Rules

- We got the following discriminant rules.
  - IF  $\text{Fe}_2\text{O}_3$ ="Low" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="Medium" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="High" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Medium" AND Rb="Medium" THEN Class="NOT C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Medium" AND Rb="Low" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Medium" AND Rb="High" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Low" AND Zn="Low" THEN Class="C<sub>1</sub>"



# More Discriminant Rules

- We got the following discriminant rules.
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND  $\text{CaO}$ ="Low" AND  $\text{Zn}$ ="Medium" AND  $\text{Tb}$ ="Low" THEN Class="NOT  $C_1$ "
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND  $\text{CaO}$ ="Low" AND  $\text{Zn}$ ="Medium" AND  $\text{Tb}$ ="Medium" THEN Class=" $C_1$ "
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND  $\text{CaO}$ ="Low" AND  $\text{Zn}$ ="Medium" AND  $\text{Tb}$ ="High" THEN Class="NOT  $C_1$ "
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND  $\text{CaO}$ ="Low" AND  $\text{Zn}$ ="High" THEN Class="NOT  $C_1$ "
- Predictive Accuracy Determined: 89.79%

# Experiment 3 Dataset 2- Results

- Experiment 3: Using Major Attributes
- Decision Trees were generated using the J48 algorithm.

```
Classifier output
J48 unpruned tree
-----

Fe2O3* = Low: C1 (34.0)
Fe2O3* = Medium: C1 (32.0)
Fe2O3* = High
|  CaO = High: C1 (4.0)
|  CaO = Medium: C1 (7.0/2.0)
|  CaO = Low
|  |  Zn = Low: C1 (2.0)
|  |  Zn = Medium: NOT C1 (8.0/1.0)
|  |  Zn = High: NOT C1 (11.0)

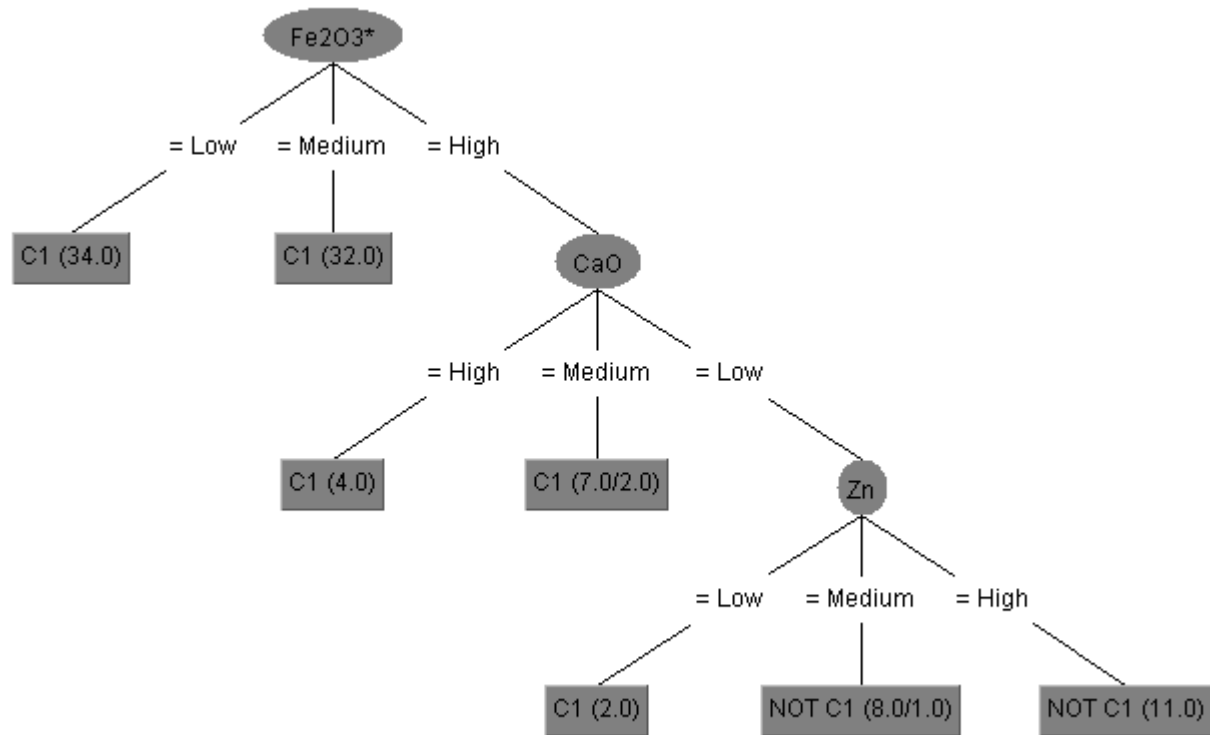
Number of Leaves :      7

Size of the tree :      10

Time taken to build model: 0 seconds

=== Stratified cross-validation ===
=== Summary ===

Correctly Classified Instances      93           94.898 %
Incorrectly Classified Instances     5           5.102 %
Kappa statistic                     0.8458
Mean absolute error                  0.0607
Root mean squared error              0.196
```





# Discriminant Rules

- We got the following discriminant rules.
  - IF  $\text{Fe}_2\text{O}_3$ ="Low" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="Medium" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="High" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Medium" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Low" AND Zn="Low" THEN Class="C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Low" AND Zn="Medium" THEN Class="NOT C<sub>1</sub>"
  - IF  $\text{Fe}_2\text{O}_3$ ="High" AND CaO="Low" AND Zn="High" THEN Class="NOT C<sub>1</sub>"

# More Discriminant Rules

- Predictive Accuracy Determined: 94.90%



# Accuracy Analysis for both Datasets

- Here is a comparison of the accuracy achieved with each Dataset.

	Dataset #1	Dataset #2
Experiment 1	70.58%	79.59%
Experiment 2	97.06%	89.79%
Experiment 3	82.35%	94.90%

- Dataset 1 was carried out using 3 bins – equal frequency discretization.
- Dataset 2 was carried out using 4 bins – equal width discretization.



# Conclusion & Thoughts

- High accuracy for a particular value can sometimes be misleading since there is a lot of data (77 records) for C<sub>1</sub> as compared to data (21 records) for other classes.
- WEKA produces different rules depending on the techniques used for data preparation.
- Dataset 2 generally had better accuracy. Thus, we can conclude that the 4-bin equal width method was slightly more accurate than the 3-bin equal frequency method.
- Comparing classes with each other gave the best overall accuracy. (i.e. comparing C<sub>1</sub> with all other classes)