Wine Classification Using Decision Trees

Weka Implementation

1. Introduction

The goal of this project was to classify different types of wines based on their chemical and physical properties. We used a dataset containing these attributes to train a decision tree model. The analysis was performed using both a custom implementation and the **Weka** software with the J48 algorithm (a derivative of C4.5). This allowed us to compare the performance of our implementation with a well-optimized tool, evaluating metrics such as accuracy, error rate, and model complexity.

2. Dataset

- Dataset Name: wine data.
- **Number of Instances:** 178 (each row represents a wine).
- **Attributes:** 14, including 13 predictive features and 1 target class.
 - o Predictive features:
 - alcohol, malic_acid, ash, alcalinity_of_ash, magnesium, total_phenols, flavanoids, nonflavanoid_phenols, proanthocyanins, color_intensity, hue, od280/od315_of_diluted_wines, proline.
 - o Target attribute: class (indicating wine type: class_0, class_1, or class_2).

3. Methodology

3.1 Implementation in Weka

- Software Used: Weka.
- **Classifier:** J48 decision tree.
- Model Configuration:
 - o Parameter -c 0.25: Confidence factor for pruning.
 - o Parameter -M 2: Minimum number of instances per leaf.
- **Evaluation Method:** 10-fold cross-validation.
- Steps Performed:
 - 1. Imported the dataset in .arff format.
 - 2. Selected the J48 algorithm in the Classify tab.
 - 3. Ran the model and analyzed results, including tree visualization and metrics.

3.2 Custom Implementation

- A custom decision tree algorithm was developed, based on information gain to select attributes.
- The model was evaluated using the same dataset and 10-fold cross-validation for consistency with the Weka experiments (default weka).

4. Results

4.1 Weka's Decision Tree Results

1. Tree Structure:

- o Number of leaves: 5.
- o Total tree size: 9 nodes.
- Key splits:
 - Main attributes: flavanoids, color_intensity, proline.
 - Example of extracted rules:
 - If flavanoids <= 1.57 and color_intensity > 3.8, then the class is class 2.
 - If flavanoids > 1.57 and proline > 720 and color intensity > 3.4, then the class is class 0.

2. Model Performance:

- o Overall Accuracy: 93.82%.
- o Error Rate: 6.18%.
- Metrics by Class:
 - class 0: True Positive Rate (TP Rate): 98.3%, Precision: 93.5%.
 - class 1: TP Rate: 94.4%, Precision: 91.8%.
 - class 2: TP Rate: 87.5%, Precision: 97.7%.
- Common Errors:
 - 1 instance of class 0 misclassified as class 1.
 - 5 instances of class 2 misclassified as class 1.

3. Confusion Matrix:

o Shows the distribution of predictions:

```
a b c <-- classified as

58 1 0 | a = class_0

3 67 1 | b = class_1

1 5 42 | c = class_2
```

4. **Efficiency:** Time to build the model: **0.05 seconds**.

4.2 Comparison with Custom Implementation

- The custom implementation achieved slightly lower performance, with an overall accuracy of **91.5%**, compared to Weka's **93.82%**.
- The generated tree in our implementation was more complex, with **12 nodes** and **7 leaves**, indicating less efficient pruning.
- Classification errors were similarly distributed but more frequent, especially for class 2.

5. Conclusions

1. Weka's Performance:

- o The J48 classifier in Weka demonstrated high efficiency and accuracy, with a compact model (9 nodes) and a precision of 93.82%.
- Classification errors were mainly concentrated in class_2, which has fewer instances in the dataset.

2. Comparison with Custom Implementation:

- The custom implementation achieved reasonable performance but was slightly less accurate (91.5%) and produced a more complex tree.
- o This indicates that Weka's pruning and parameter optimization processes are more effective.

3. Key Takeaways:

 Implementing algorithms manually provides a deep understanding of their functionality, but optimized tools like Weka offer superior performance and more efficient designs.

6. Appendices

6.1 Tree Generated by Weka

