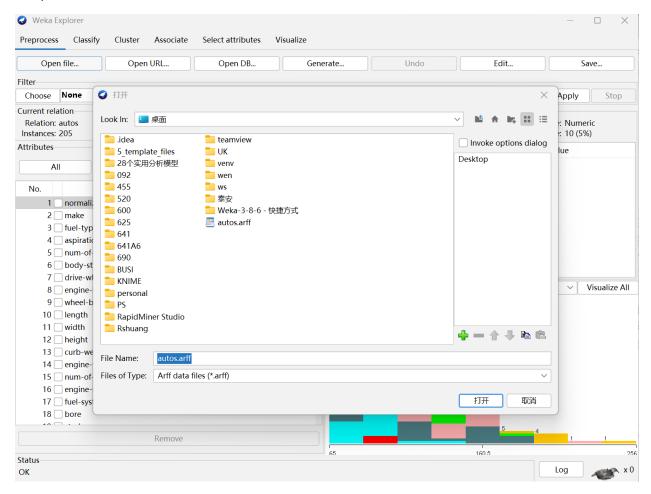
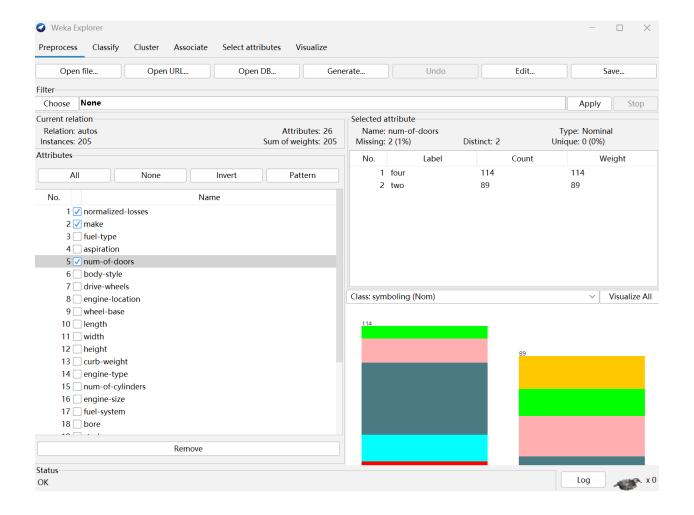
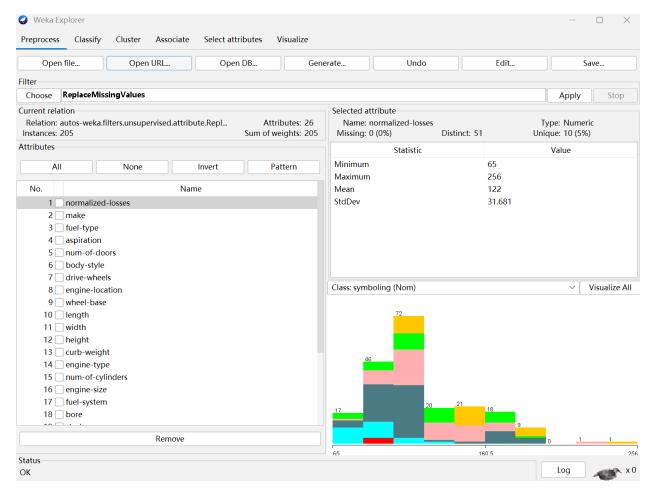
WEKA Challenge

After studying, I started my WEKA challenge. The first challenge I encountered was finding the ARFF files, which I had never used before. I searched and downloaded the dataset for my challenge (autos) from OpenML (an open machine learning platform). The dataset consists of three types of entities: (a) specifications of various features of the car, (b) specified insurance Risk Rating, (c) Normalized loss of use compared to other cars.

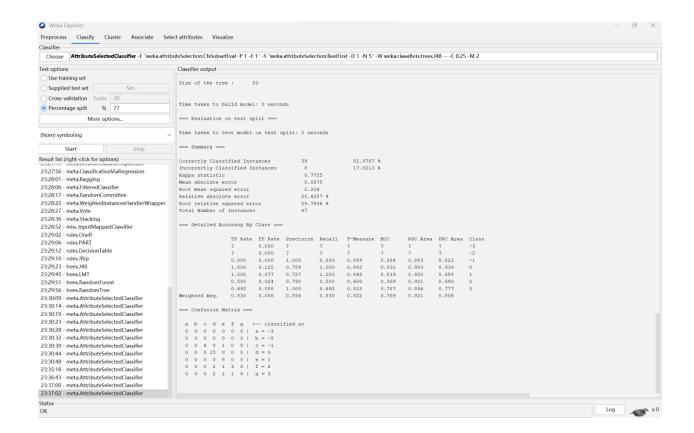


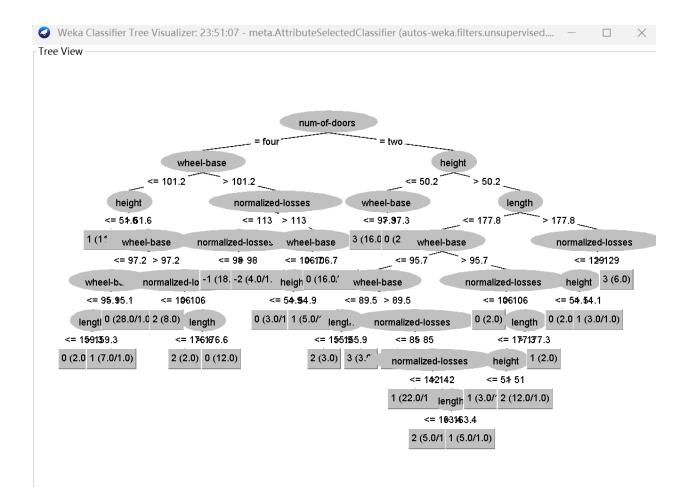
After loading the document into WEKA I need to perform data processing first and we can see that the file has missing values. We can choose unsupervised attributes to replace missing values. ReplaceMissingValues provides a variety of methods to deal with missing values, including using mean, median, mode, fixed value or custom strategy. This flexibility helps choose the most suitable data Alternative methods for gathering and analyzing requirements. The benefit of this is that it helps maintain data integrity, ensuring that every instance in the data set contains complete information.



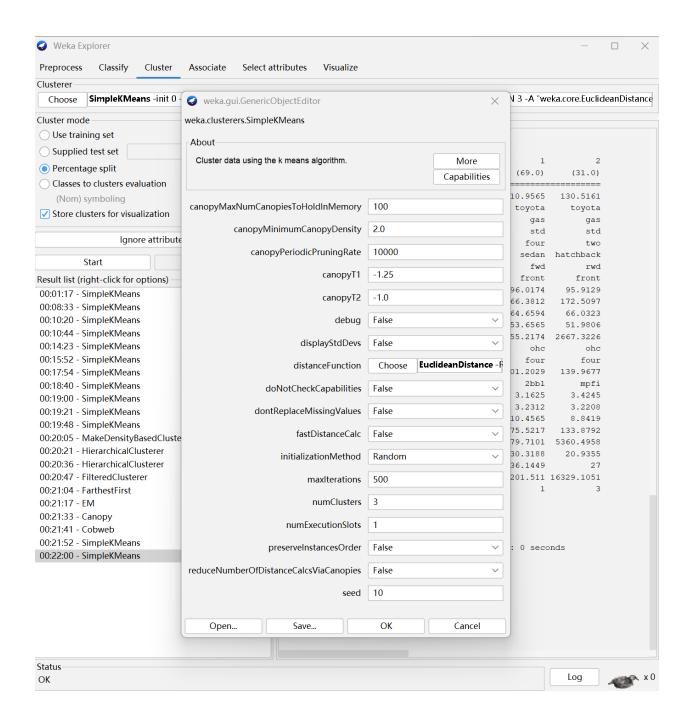


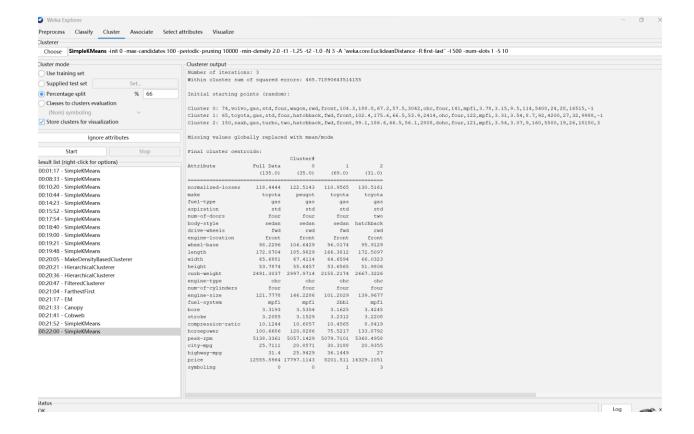
Then I wanted to classify the data. I tried many methods, but the percentage of Correctly Classified Instances was very low. As you can see from the picture, I finally chose AttributeSelectedClassifier for classification. This classification model uses a composite approach, including feature selection and J48 decision trees, to classify the test data and performs well overall. The model selected attributes with predictive capabilities through feature selection, and then used decision trees for classification. It achieved a correct classification rate of about 83%, and the Kappa statistic was 0.7755, indicating that the model has good performance. The confusion matrix provides classification results for different categories and can be used to evaluate the performance of the model in detail. In addition, we can also choose the visualization function to visualize the classification model. We can see the decision tree model and explore the rules and branches of different nodes, which can help us understand how the model classifies data.





I then wanted to cluster this data, and I chose the k-Means clustering algorithm because it works well with large data sets and is usually able to complete the clustering task in a relatively short time. From the clustering results, the data set is divided into three different clusters (Cluster 0, Cluster 1, Cluster 2), each cluster contains a different number of instances. For each cluster, you see the average values of the individual attributes (Final cluster centroids), and these values reflect the characteristics of each cluster. For example, you can compare attributes such as average vehicle price, engine size, and horsepower among different clusters. By analyzing the average value of each cluster and the composition of vehicle models within the cluster, we can identify similar car groups, which can help with decision-making and insights in application areas such as market positioning, product classification, and customer analysis. However, looking at WCSS, its number is larger, which means that the data points in the clusters are not close enough. According to the visualization, it can also be seen that each cluster is not close enough. I also tried other clustering algorithms, but none of them performed very well.







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