A decorative graphic on the left side of the slide consisting of two overlapping parallelograms. The front one is blue and the back one is a light greenish-blue. Both are tilted at an angle.

A Comparative Study on Car Evaluation dataset using different machine-learning methods

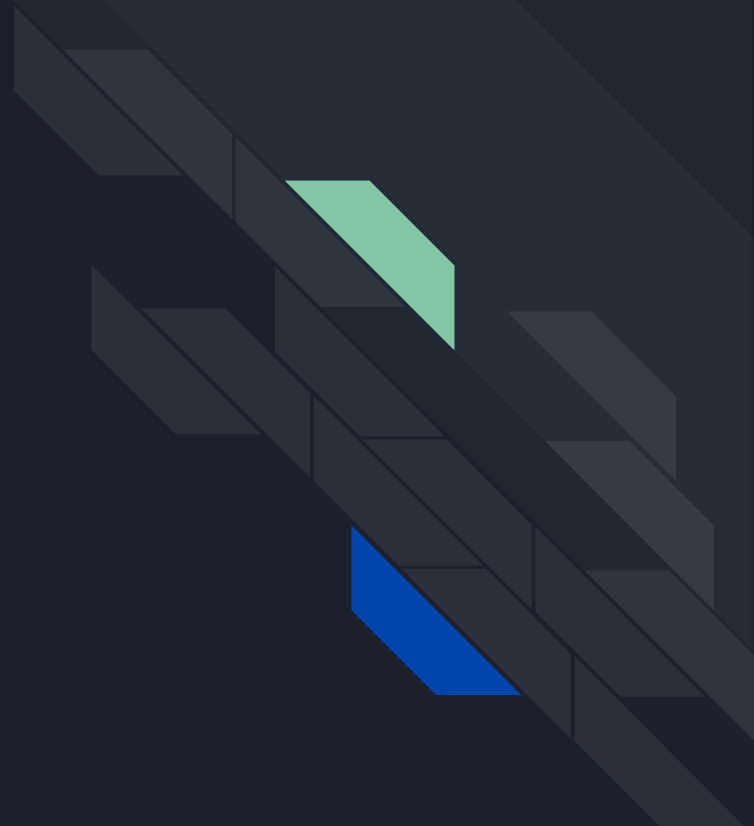
Group member: Shiyu Wang, Yingda Tao, Hao Zeng, Yuqing Jin



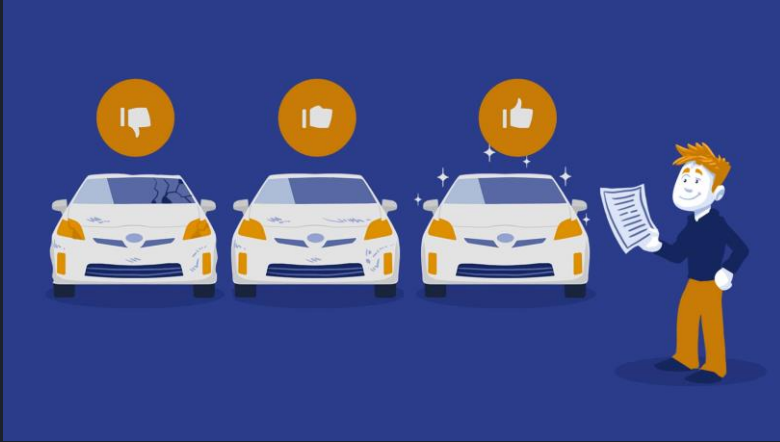
Body of Representation

- Dataset Description
- Reproduction (NB & MLP-ANN)
- Other ML methods (SVM, KNN, Decision Tree, Random Forest, etc.)
- Conclusion

Dataset Description



Data Description

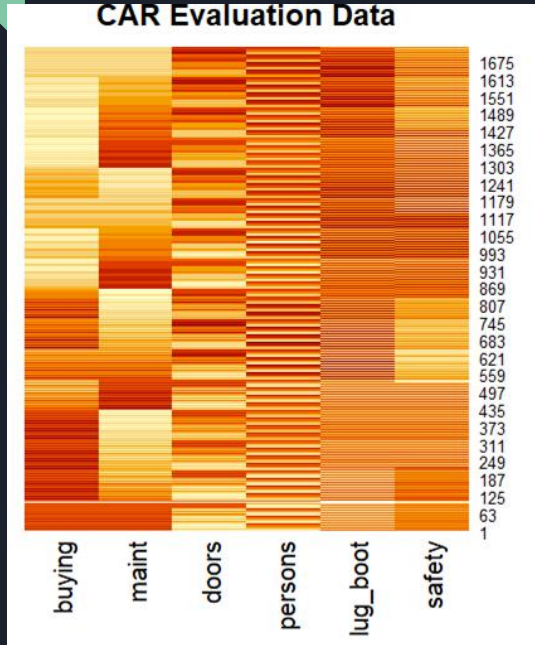


People would make an evaluation before buying a car. They would consider the price, maintenance, seat number, door number and luggage boot etc.

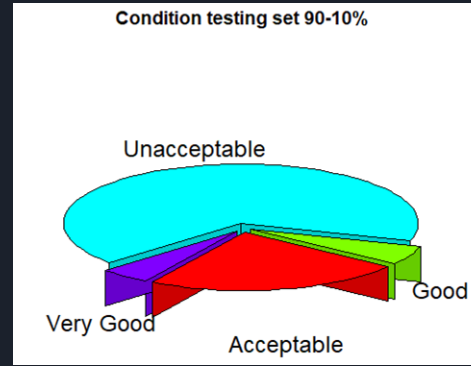
Number of Instances:	1728
Attribute 1: Buying	vhigh, high, med, low
Attribute 2: Maintenance	vhigh, high, med, low
Attribute 3: Doors	2, 3, 4, 5more
Attribute 4: Persons	2, 4, more
Attribute 5: Luggage boot	small, med, big
Attribute 6: Safety	low, med, high
Decision Attribute: Class	unacc, acc, good, vgood

Six predictors and one response in car evaluation dataset.
Total 1728 instances in the dataset.

Data Visualization



Heatmap (clustered heatmap) of our six attributes. From the graph, we could see how dispersed each of attribute is.



Pie chart of the decision attribute (class). Most of the instances belong to 'unacceptable'. 'Good' and 'very good' are far less than others.



Correlation matrix of all the attributes. Correlation coefficients are denoted on the graph.

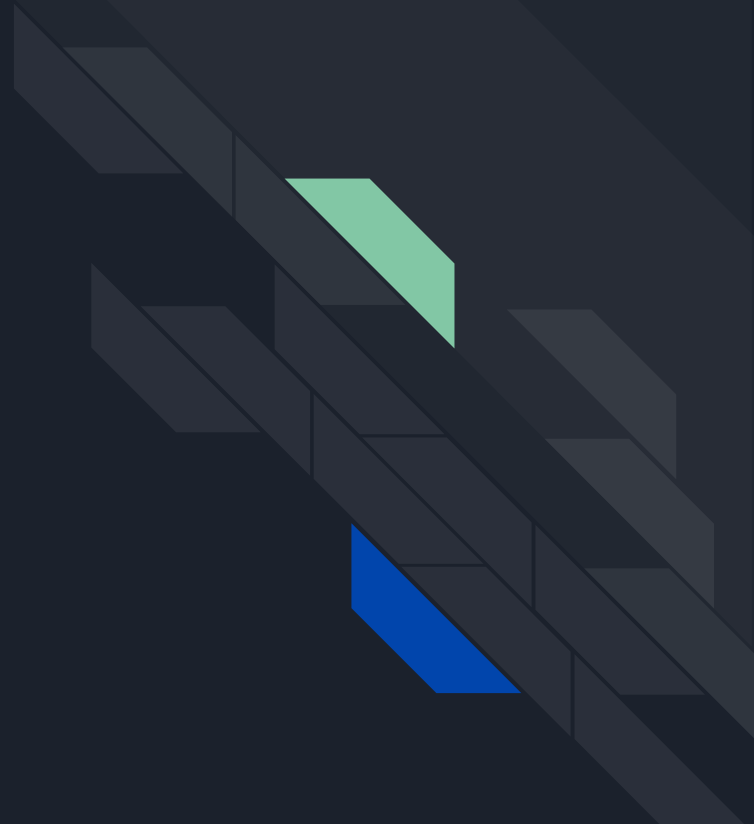
Data Preprocessing

- **Data-Cleaning:**
Convert nominal attributes to numerical value
- **Min-Max Normalization:**
Narrow data between 0-1
- **Data-Split:**
Training : Testing =
90%-10%, 66%-34%, 50%-50%,
and 10-fold cross validation

Attribute	Nominal	New numerical value	Attribute	Nominal	New numerical value
Buying	vhigh	4	Persons	2	2
	high	3		4	4
	med	2		more	5
	low	1	Luggage boot	small	3
Maintenance	vhigh	4		med	2
	high	3		big	1
	med	2	Safety	high	3
	low	1		med	2
Doors	2	2		low	1
	3	3	Class	vgood	3
	4	4		good	2
	5more	5		acc	1
				unacc	0

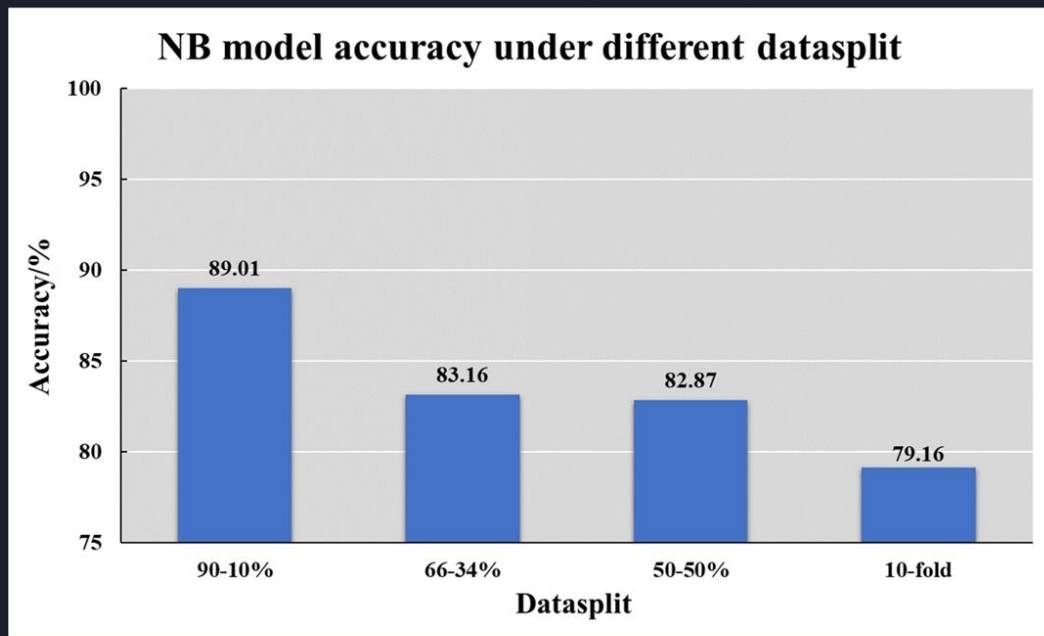
Change 6 predictors and 1 decision attribute to numerical value, to facilitate data analysis.

Reproduction



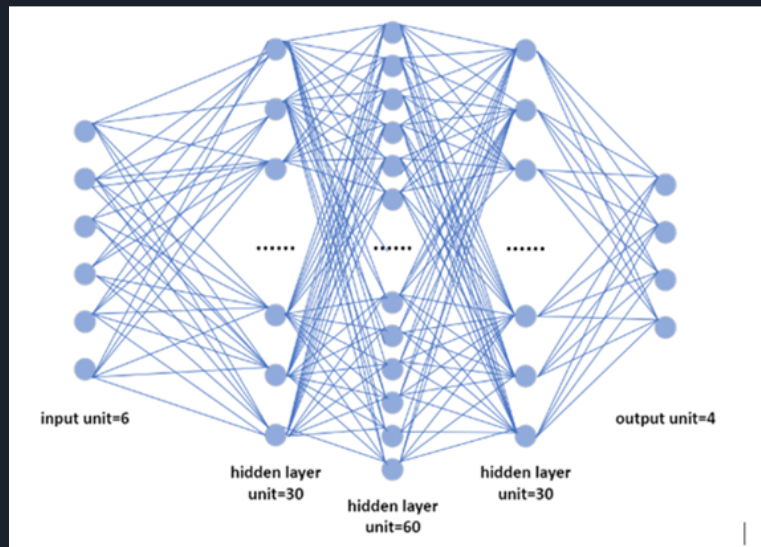
Naive Bayes

$$\hat{f}(x) = \operatorname{argmax}_{y \in Y} P[X = x|Y = y] * P[Y = y] = \operatorname{argmax}_{y \in Y} \prod_{j=1}^d P[X_j = x_j|Y = y] * P[Y = y]$$



The structure of ANN

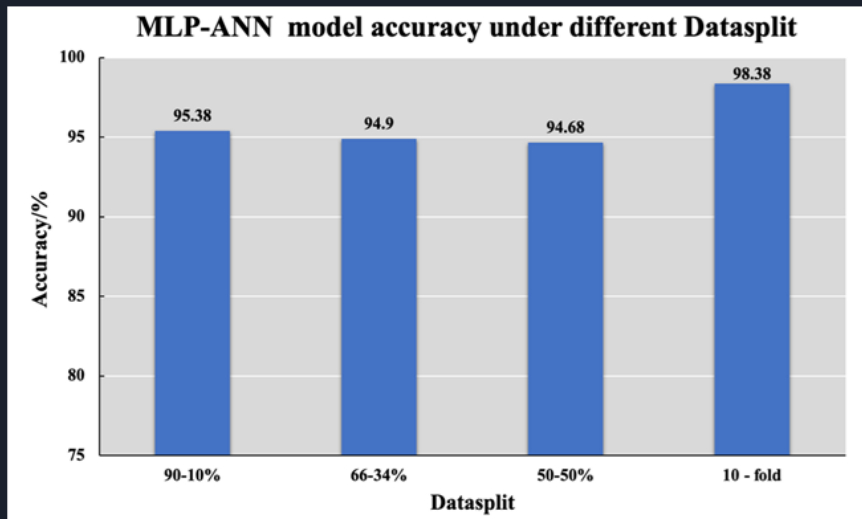
MLP-ANN



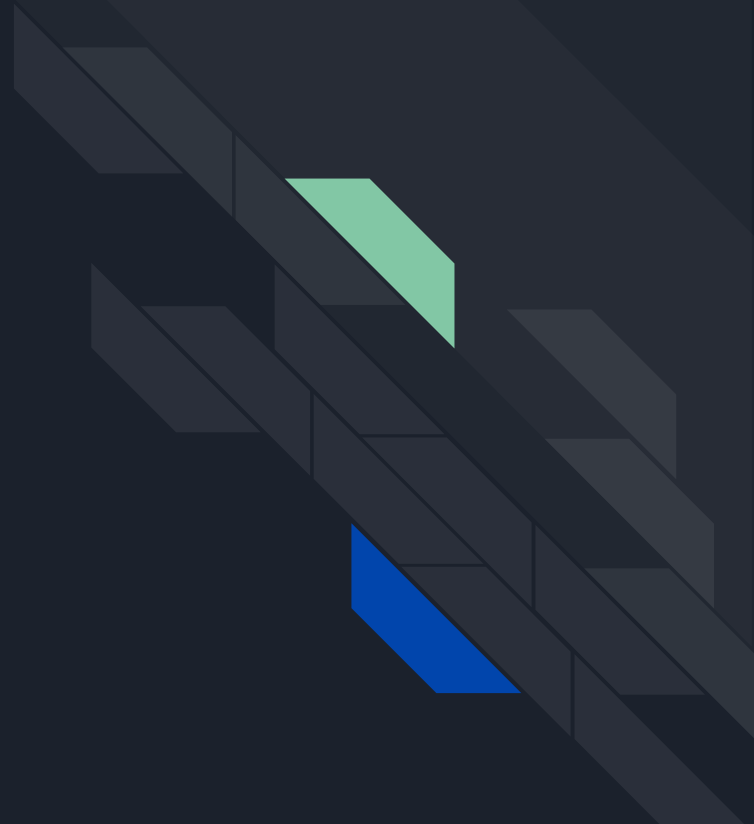
The structure of ANN

NB model accuracy in the original paper

Data Split	Accuracy
90-10%	94.79%
66-34%	93.19%
50-50%	92.70%
10-fold	94.09%

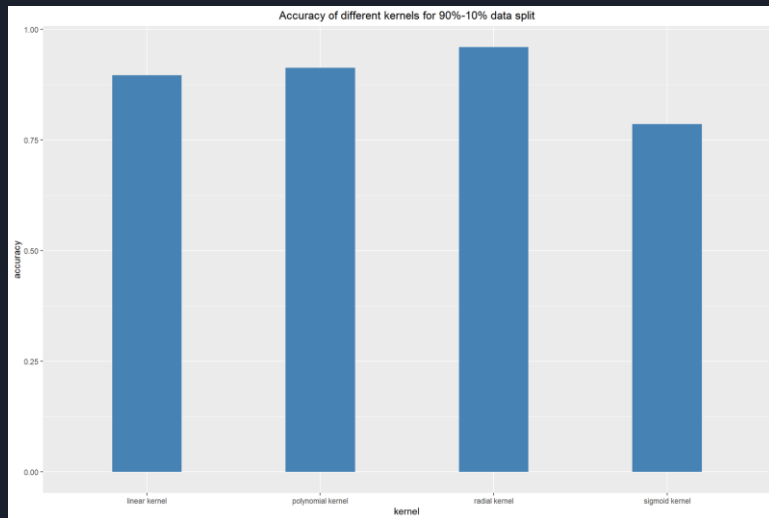


Other ML methods

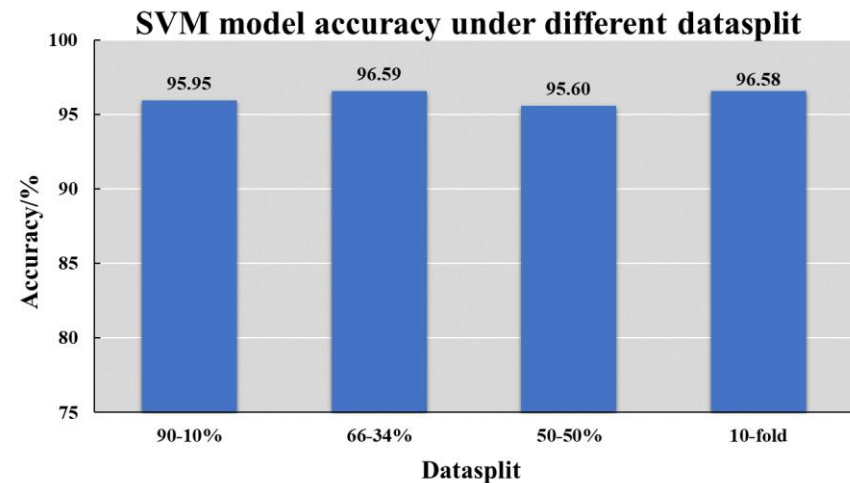


SVM_[3]

SVM using kernel trick

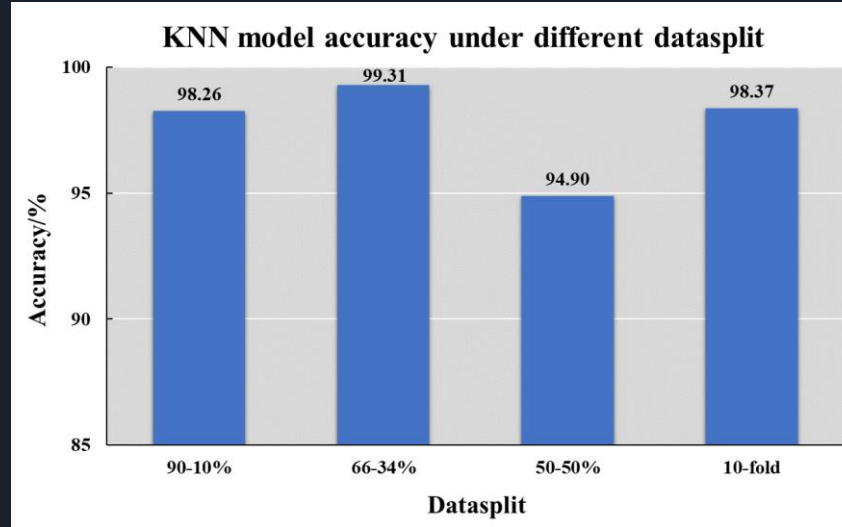


Accuracy of SVM model using radial kernel



KNN_[4]

Accuracy of KNN model with K=5



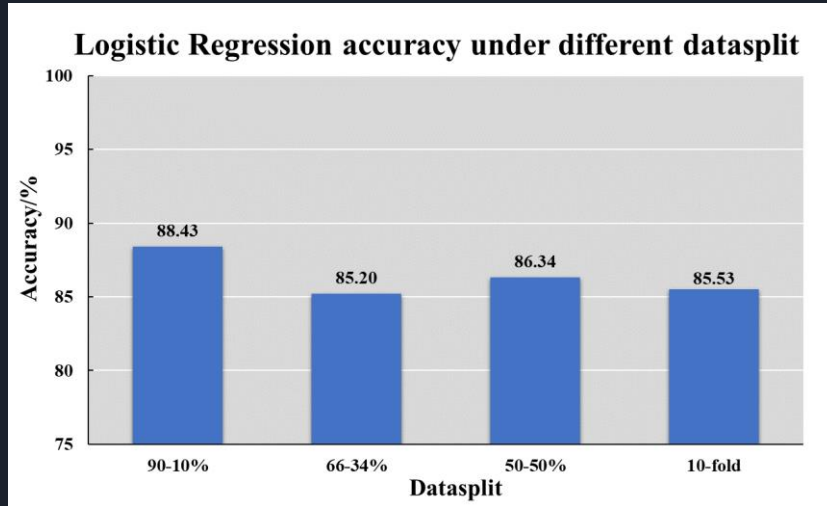
When k is too small \rightarrow sensitive to noise (overfitting problem)

When k is too large \rightarrow bad accuracy ($k=200$, accuracy decreases to 80%!)

Logistic Regression

Not regression method, but a classification method [5]

Accuracy of Logistic Regression

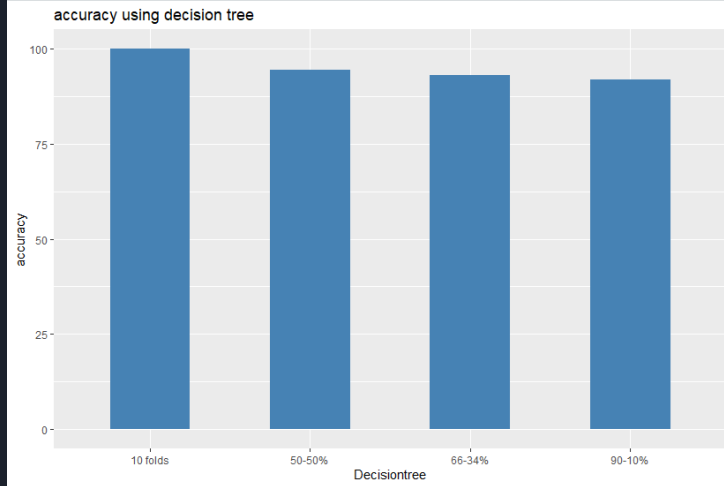


Result is not good: Logistic Regression method is usually used for solving binary classification problems, not suitable for multi-classification problem.

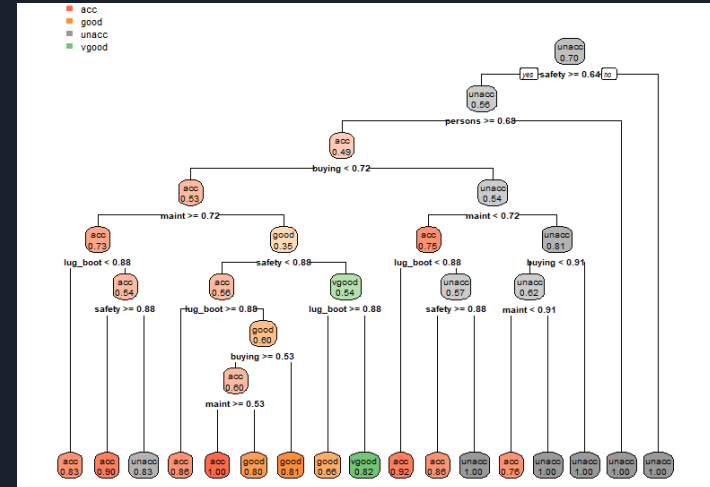
Decision Tree Classification^[1]

- Decision tree is a nonparametric method.
- It can be used to compare with parametric methods (such as naive Bayes)
- CART algorithm

Accuracy of decision tree method



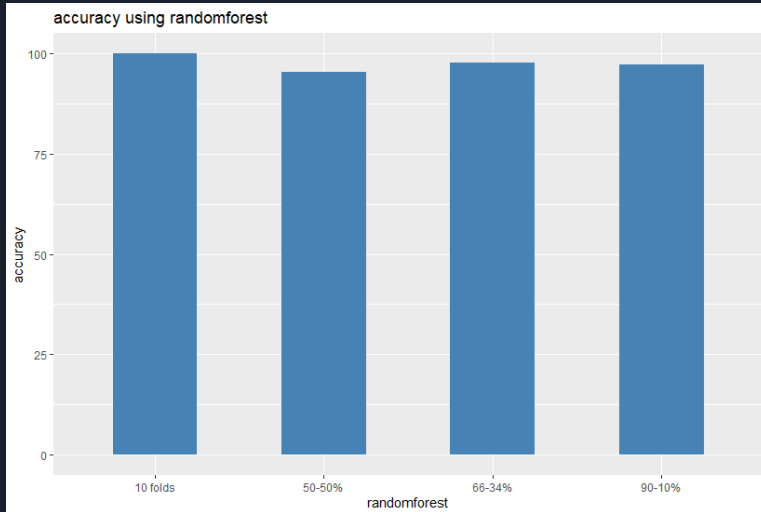
visualization of decision tree



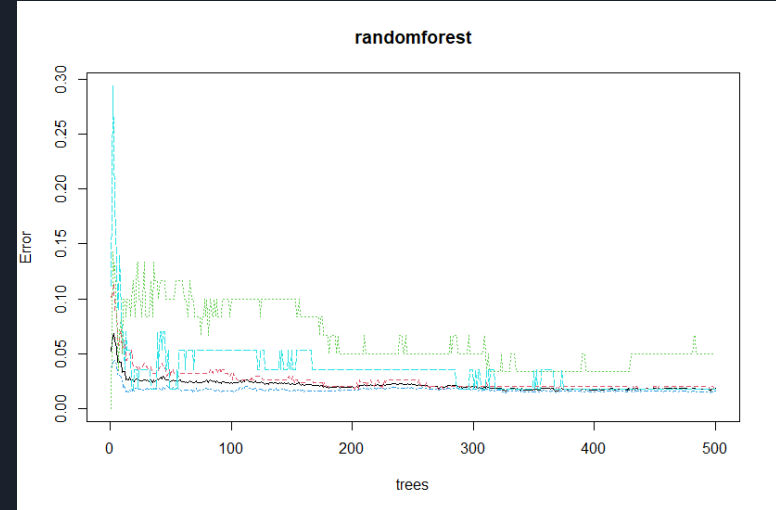
Randomforest[2]

- Random forest is composed of many decision trees(500 trees)
- A random forest has higher accuracy for data processing than a decision tree

Accuracy of random forest method



Randomforest 500 tree error model



Conclusion

- Only Naïve Bayes and Logistic Regression accuracy below 94%
- KNN performed the best whether in terms of accuracy and efficiency.
- Preprocess the data more suitable may help for the accuracy(LOOCV, Boosting and Adaboost)

Performance of each method

Method	Average Accuracy	Average Running Time/s
MLPNN	95.84%	10.69
Naïve Bayes	83.55%	0.59
SVM	95.78%	0.05
KNN	97.71%	0.02
Logistic Regression	86.35%	0.25
Decision Tree	94.71%	0.38
Random Forest	97.46%	0.17

Reference

- [1] Gupta, P. (2017, November 12). Decision trees in machine learning. Medium. Retrieved December 19, 2021
- [2] Maklin, C. (2019, July 30). Random Forest in R. Medium. Retrieved December 19, 2021
- [3] Hearst, M. A., Dumais, S. T., Osuna, E., Platt, J., & Scholkopf, B. (1998). Support vector machines. *IEEE Intelligent Systems and their applications*, 13(4), 18-28.
- [4] Guo, G., Wang, H., Bell, D., Bi, Y., & Greer, K. (2003, November). KNN model-based approach in classification. In "OTM Confederated International Conferences" On the Move to Meaningful Internet Systems" (pp. 986-996). Springer, Berlin, Heidelberg.
- [5] de Souza, R. M., Cysneiros, F. J. A., Queiroz, D. C., & Roberta, A. D. A. (2008, October). A multi-class logistic regression model for interval data. In 2008 IEEE International Conference on Systems, Man and Cybernetics (pp. 1253-1258). IEEE.

Thanks for your listening!

The background features a series of dark gray, three-dimensional rectangular blocks arranged in a perspective view, receding towards the right. A light green parallelogram is positioned above a blue parallelogram, both of which are also in perspective and appear to be part of the geometric structure.