

PREDICTING USED CAR PRICE TYPE

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Abstract

The number of cars on Mauritian roads has been rising consistently by 5% during the last decade. In 2014, 173 954 cars were registered at the National Transport Authority. Thus, one Mauritian in every six owns a car, most of which are second hand reconditioned cars and used cars. The aim of this study is to assess whether it is possible to predict the price of second-hand cars using artificial neural networks. Thus, data for 200 cars from different sources was gathered and fed to four different machine learning algorithms. We found that support vector machine regression produced slightly better results than using a neural network or linear regression. However, some of the predicted values are quite far away from the actual prices, especially for higher priced cars. Thus, more investigations with a larger data set are required and more experimentation with different network type and structures is still required in order to obtain better predictions.

1. INTRODUCTION

According to the data obtained from the National Transport Authority (2014), there has been an increase of 254% in the number of cars from 2003 (68, 524) to 2014 (173, 954), as shown in Figure 1. We can thus infer that the sale of second-hand imported (reconditioned) cars and second-hand used cars has eventually increase given that new cars represent only a very small percentage of the total number of cars sold each year. Most individuals in Mauritius who buy new cars also want to know about the resale value of their cars after some years so that they can sell it in the used car market.

Price prediction of second-hand cars depends on numerous factors. The most important ones are manufacturing year, make, model, mileage, horsepower and country of origin. Some other factors are type and amount of fuel per usage, the type of braking system, its acceleration, the interior style, its physical state, volume of

cylinders (measured in cubic centimeters), size of the car, number of doors, weight of the car, consumer reviews, paint colour and type, transmission type, whether it is a sports car, sound system, cosmic wheels, power steering, air conditioner, GPS navigator, safety index etc. In the Mauritian context, there are some special factors that are also usually considered such as who were the previous owners and whether the car has had any serious accidents.

Thus, predicting the price of second-hand cars is a very laudable enterprise. In this paper, we will assess whether neural networks can be used to accurately predict the price of secondhand cars. The results will also be compared with other methods like linear regression and support vector regression.

This paper proceeds as follows. In this system, various works on neural networks and price prediction have been summarized. The methodology and data



collection are described in this system. The system presents the results for price prediction of second-hand cars. Finally, we end the paper with a conclusion and some ideas towards future works.

2. EXISTING SYSTEM

Predicting the price of second-hand cards has not received much attention from academia despite its huge importance for the society. Bharambe and Dharmadhikari (2015) used artificial neural networks (ANN) to analyse the stock market and predict market behaviour. They claimed that their proposed approach is more accurate than existing ones by 25%.

Pudaruth (2014) used four different supervised machine learning techniques namely kNN (k-Nearest Neighbour), Naïve Bayes, linear regression and decision trees to

predict the price of second-hand cars. The best result was obtained using kNN which had a mean error of 27000 rupees.

Jassbi et al. (2011) used two different neural networks and regression methods to predict the thickness of paint coatings on cars. The error for the final thickness of the paint was found to be 2/99 microns for neural networks and 17/86 for regression. Ahangar et al. (2010) also compared the use of neural networks with linear regression in order to predict the stock prices of companies in Iran. They also found that neural networks had superior performance both in terms of accuracy and speed compared to linear regression. Listiani (2009) used support vector machines (SVM) to predict the price of leased cars.

They showed that SVM performed better than simple linear regression and

multivariate regression. Iseri and Karlik (2009) used neural networks to predict the price of automobiles and achieved a mean square error of 8% compared with 14.4% for regression. Yeo (2009) used neural networks to predict the retention rate for policy holders of automobile insurance. The neural network was able to predict which customers were likely to renew their policy and which ones would terminate soon. Doganis et al. (2006) used artificial neural networks and genetic algorithm in order to predict the sales of fresh milk with an accuracy of 95.4%. Rose (2003) used neural networks to predict the production of cars for different manufacturers.

Disadvantages

An existing methodology doesn't implement DATA PRE-PROCESSING & LABELLING method.

The system not implemented an effective ML Classifiers for predictions in the datasets.

3. PROPOSED SYSTEM

In order to carry out this study, data have been obtained from different car websites and from the small adverts sections found in daily newspapers like L'Express and Le Defi. The data was collected in less than one month interval (i.e. in the month of August in 2014) because like other goods, the price of cars also changes with time. Two hundred records were collected. The data comprises of different features for second-hand cars such as the year (YEAR) in which it was manufactured, the make (MAKE), engine capacity (ENGINE) measured in cubic centimetres, paint (PAINT) type (normal or metallic), transmission (T/N) type (manual or automatic), mileage (MILEAGE) (number

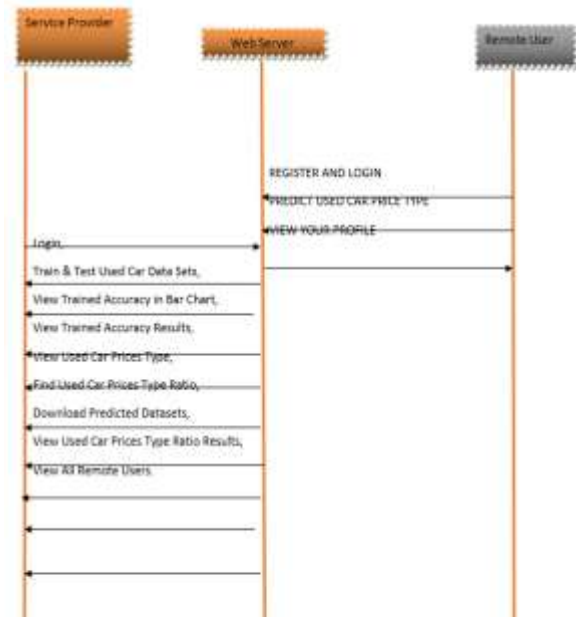
of kilometres the car has been driven) and its price (PRICE) in Mauritian rupees. A large number of experiments have been conducted in order to find the best network structure and the best parameters for the neural network. We found that a neural network with 1 hidden layer and 2 nodes produced the smallest mean absolute error among various neural network structures that were experimented with. However, we found that Support Vector Regression and a multilayer perception with back-propagation produced slightly better predictions than linear regression while the k-Nearest Neighbour algorithm had the worst accuracy among these four approaches. All experiments were performed with a cross validation value of 10 folds.

Advantages

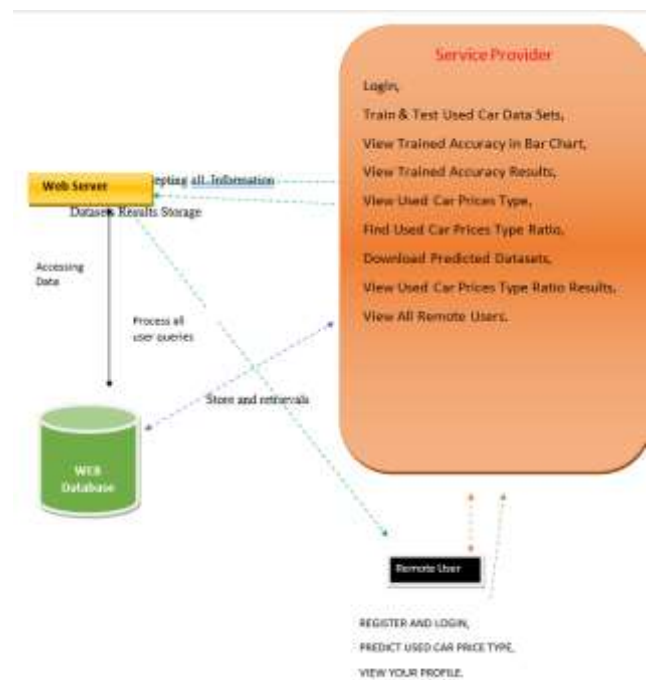
The purpose of linear regression, support vector regression which are more effective for testing and training accuracy.

In this work, the system will assess whether neural networks can be used to accurately predict the price of secondhand cars.

Sequence Diagram



4. ARCHITECTURE DIAGRAM



5. CONCLUSION

The aim of this paper was to predict the price of second-hand reconditioned and second-hand used cars in Mauritius. The car market has been increasing steadily by



around 5% for the last ten years, showing the high demand for cars by the Mauritian population. There are hundreds of car websites in Mauritius but none of them provide such a facility to predict the price of used cars based on their attributes. Our dataset of 200 records was used with the cross-validation technique with ten folds. The car make, year manufactured, paint type, transmission type, engine capacity and mileage have been used to predict the price of second-hand cars using four different machine learning algorithms. The average residual value was reasonably low for all four approaches. Thus, we conclude that predicting the price of second-hand cars is a very risky enterprise but which is feasible. This system will be very useful to car dealers and car owners who need to assess the value of their cars. In the future, we intend to collect more data and more features and to use a larger variety of machine learning algorithms to do the prediction.

6. REFERENCES

- [1] NATIONAL TRANSPORT AUTHORITY. 2015. Available at: <http://nta.govmu.org/English/Statistics/Pages/Archives.aspx>. [Accessed 24 April 2015].
- [2] Bharambe, M. M. P., and Dharmadhikari, S. C. (2015) "Stock Market Analysis Based on Artificial Neural Network with Big data". *Fourth Post Graduate Conference, 24-25th March 2015, Pune, India*.
- [3] Pudaruth, S. (2014) "Predicting the Price of Used Cars using Machine Learning Techniques". *International Journal of Information & Computation Technology*, Vol. 4, No. 7, pp.753- 764.
- [4] Jassibi, J., Alborzi, M. and Ghoreshi, F. (2011) "Car Paint Thickness Control using Artificial Neural Network and Regression Method". *Journal of Industrial Engineering International*, Vol. 7, No. 14, pp. 1-6, November 2010
- [5] Ahangar, R. G., Mahmood and Y., Hassen P.M. (2010) "The Comparison of Methods, Artificial Network with Linear Regression using Specific Variables for Prediction Stock Prices in Tehran Stock Exchange". *International Journal of Computer Science and Information Security*, Vol.7, No. 2, pp. 38-46.
- [6] Listiani, M. (2009) "Support Vector Regression Analysis for Price Prediction in a Car Leasing Application". Thesis (MSc). Hamburg University of Technology.
- [7] Iseri, A. and Karlik, B. (2009) "An Artificial Neural Network Approach on Automobile Pricing". *Expert Systems with Application: ScienceDirect Journal of Informatics*, Vol. 36, pp. 155-2160, March 2009.
- [8] Yeo, C. A. (2009) "Neural Networks for Automobile Insurance Pricing". *Encyclopedia of Information Science and Technology*, 2nd Edition,