IST 707 Applied Machine Learning By Prof. Kelvin King

Assignment 5

Decision Tree

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Authorship Attribution Report

Section 1: Data Preparation

In this report, we aim to attribute the correct authorship of the disputed essays from Federalist papers. The dataset was first split using train_test_split method into training and testing data. The dataset has total of 11 disputed authorship which we try to identify the right author using our analysis.

The training and testing data was in ratio of 80/20.

The feature set for our classification model was based on the frequency distribution of common function words in the essays.

Section 2: Decision Tree Model

Decision Tree Model

We built a decision tree classification model using the default settings. The decision tree model was trained on the training dataset, and we evaluated its performance on the testing dataset.

Section 3: Prediction

After training the decision tree model, we used them to predict the authorship of the disputed essays. The predictions for the disputed essays are as follows:

Disputed Essay 1: Author: Madison
Disputed Essay 2: Author: Madison
Disputed Essay 3: Author: Jay
Disputed Essay 4: Author: Madison
Disputed Essay 5: Author: Madison
Disputed Essay 6: Author: Madison
Disputed Essay 7: Author: Madison
Disputed Essay 8: Author: Madison
Disputed Essay 9: Author: Madison
Disputed Essay 10: Author: Madison
Disputed Essay 11: Author: Madison

Model Evaluation Metrics

Conclusion

With an accuracy of approximately 93.33% and a precision of 88.88%, the model successfully identified James Madison as the likely author of the disputed essays.

The recall score of 91.6% indicates that the model correctly identified a significant portion of Madison's essays.

In conclusion, the decision tree model predicts that James Madison is the probable author of the disputed essays. These results are consistent with the findings of previous clustering methods.

Assignment 2 Report: Bike Rental Predictions with Decision Trees

Section 1: Data Preparation

In this section, we separated the original dataset into training and testing data for classification experiments. We used two datasets, hour.csv and day.csv, for predicting hourly and daily rental counts, respectively. The following splits were made:

Hourly Rental Count Model:

Training Data: 80% of hourly data Testing Data: 20% of hourly data **Daily Rental Count Model:** Training Data: 80% of daily data Testing Data: 20% of daily data

Section 2: Build and Tune Decision Tree Models

Built decision tree models using default settings and then tuned the parameters to evaluate if better models could be generated. Based on the max_depth and min_samples_leaf hyperparameters.

Hourly Rental Count Model - Default Settings:

Mean Squared Error (MSE): 3392.90

R-squared (R^2): 0.8929

Daily Rental Count Model - Default Settings:

Mean Squared Error (MSE): 968219.27

R-squared (R^2): 0.7585

Hourly Rental Count Model - Tuned Settings:

Best Hyperparameters: {'max depth': None, 'min samples leaf': 4}

Mean Squared Error (MSE): 2778.51

R-squared (R^2): 0.9123

Daily Rental Count Model - Tuned Settings:

Best Hyperparameters: {'max depth': None, 'min samples leaf': 8}

Mean Squared Error (MSE): 763530.46

R-squared (R^2): 0.8096

Best hyperparameters were searched using grid search.

Section 3: Prediction

In this section, we observed the results and compared the models with default and tuned settings. The following are the insights noticed:

The default hourly rental count model performed well with an R-squared value of 0.8929, indicating a good fit.

The daily rental count model with default settings had a lower R-squared value of 0.7585, which suggests that the model might not capture the underlying patterns effectively.

The hourly rental count model improved significantly after hyperparameter tuning with an R-squared value of 0.9123 and a lower MSE. The best hyperparameters were found to be max_depth=None and min_samples_leaf=4 using grid search.

Similarly, the daily rental count model improved after tuning, with an R-squared value of 0.8096 and a lower MSE. The best hyperparameters were max_depth=None and min_samples_leaf=8.

Conclusion

After tuning the decision tree models with hyperparameters, the models showed better results significantly for predicting bike rental counts. By choosing the right hyperparameters, model predictive accuracy increases.

I certify that this assignment represents my work. I have not used any unauthorized or unacknowledged assistance or sources in completing it, including free or commercial systems or services offered on the internet.

Reference:

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