Churn Prediction

Case Study Galvanize

Overview

 A ride-sharing company in San-Francisco is interested in predicting their rider retention and make data based business decision

- This case study helps the company to achieve the following goals:
 - Analyze their user data
 - Build a churn prediction model
 - Analyze the business impact
 - Make recommendations



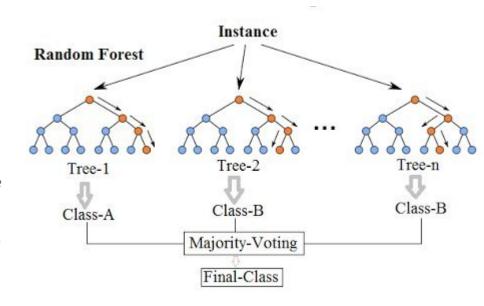


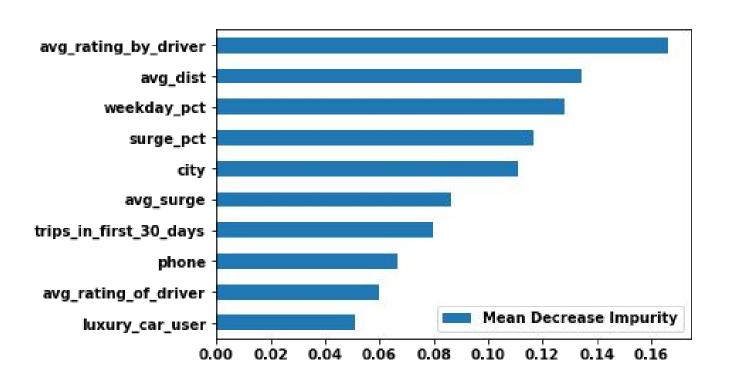
Feature Engineering

- There are 40,000 user activity items, and each item has 14 features. Based on the exploratory data analysis, we did the following feature engineering.
 - Create a "Churn" Feature
 - Conversion the signup_date to datetime object, create a churn rate features
 - Dealing with missing values.
 - Convert categorical column into numeric labels
 - Fill in the nan values for numeric column with the median value of each column
- Before model building, we split the raw data sets into train/test by 80:20. The test data was only used to evaluate the model in the last step.

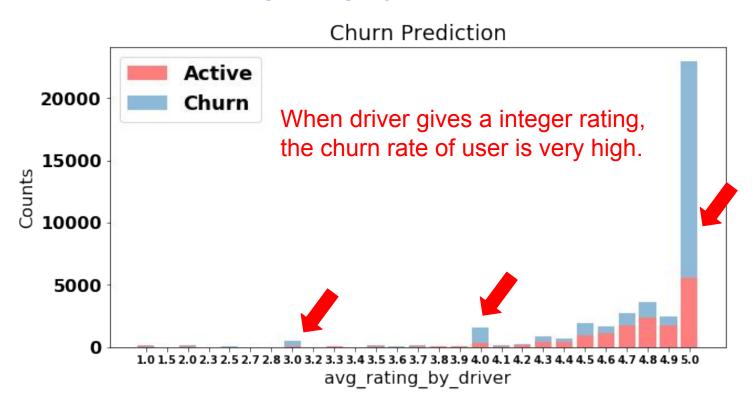
Model Selection: Random Forest

- Random Forest has several advantages for this project
 - Flexibility to handle both numeric and categorical features
 - Easy interpretation and fast training
 - Lower risk of overfitting and less variance
 - Feature importance output
- Cross-validation and grid search was used to optimize the model parameters

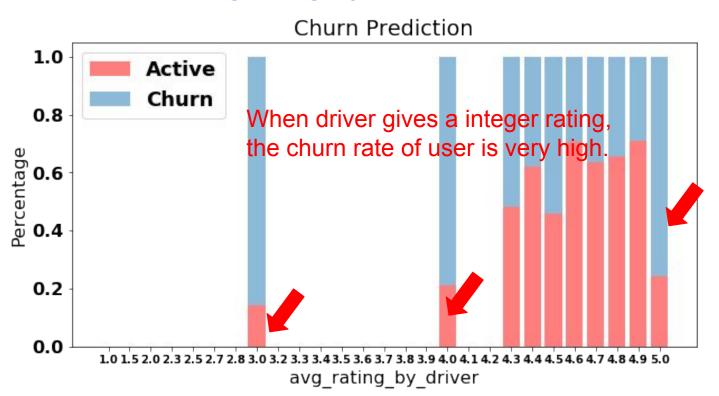


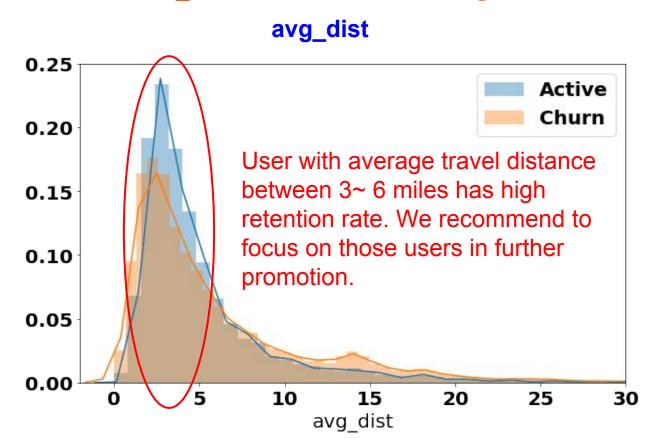


avg_rating_by_driver



avg_rating_by_driver





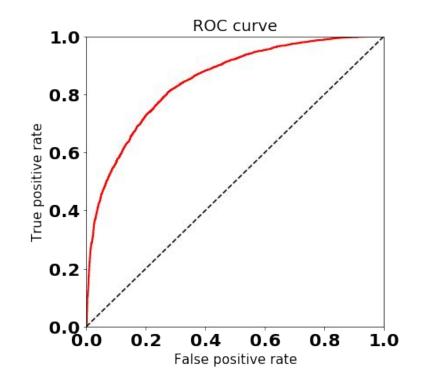
Model Prediction Performance

Accuracy Score: 0.779 Precision: 0.85

Recall: 0.80

Confusion Matrix on Test Data

| | Condition Positive | Condition Negative |
|-----------------------|-----------------------|-----------------------|
| Predicted Positive | 5318 | 910 |
| Predicted Negative | 1322 | 2450 |



Business Impact

Cost/Benefit Matrix

| | Condition Positive | Condition Negative |
|-----------------------|-----------------------|-----------------------|
| Predicted Positive | -20 \$ | 0 |
| Predicted Negative | -100\$ | 0 |

Confusion Matrix on Test Data

| | Condition Positive | Condition Negative |
|-----------------------|-----------------------|-----------------------|
| Predicted Positive | 5318 | 910 |
| Predicted Negative | 1322 | 2450 |

- Business Information: Assume the company has 10,000 customer. If one customer churns, the company lose \$100 in CLTV revenue. When we predict a customer churn, we recommend to send out a 10\$ coupon, and assume 90% of customer stays. The cost/benefit matrix was shown as above.
- Based on the prediction performance of our model, the data-driven action based on our random forest model is going to save the company 394,000\$ CLTV revenue.

Recommendation

- Based on feature analysis, we recommend to promote the driver engagement. The more detail the driver rate the rider, the lower the churn rate of the user.
- The customer service should be more focused on users who has average distance between 3 [~] 6 miles
- When our model predict a user has a high chance of churning, send out a promotion coupon to keep them.