Customer Churn Prediction

Customer Relationship Management (CRM) is a comprehensive strategy for building, managing and strengthening loyal and long-lasting customer relationships. It is broadly acknowledged and extensively applied to different fields, for example, telecommunications, banking and insurance, retail market, etc.

In any business strategy, one of its main objectives is customer retention. For example, Vodafone should asses the customers' activities (from their call logs) and hence predict the customers' churn rate, so that a measure (a special incentive to customers) can be taken to retain them accordingly. The importance of this objective is obvious, given the fact that the cost for customer acquisition is much greater than the cost of customer retention (in some cases it is 20 times more expensive).

Thus, tools to develop and apply customer retention models (churn models) are required and are essential Business Analytics applications. In the dynamic market environment, churning could be the result of low-level customer satisfaction, aggressive competitive strategies, new products, regulations, etc. Churn models aim to identify early churn signals and recognize customers with an increased likelihood to leave voluntarily.

A customer call data with 5000 records is known to us related to the customers' call logs of a mobile phone service provider. We call this data as Customer Churn Data (and abbreviated as CCD for brevity). The data set is synthetic and based on claims similar to a real world service provider. The meta data of the CCD data is as follows.

Variable Account_length Total_eve_charge area code Num total_night_minutes international_plan total_night_calls voice_mail_plan total_night_charge number_vmail_messages total_intl_minutes total_day_minutes total_intl_calls	Name (number of months active user) (total charge of evening calls) (area code of customer) (total minutes of night calls) (local/international call) (total number of night calls) (voice mail or normal) (total charge of night calls) (number of voice-mail messages) (total minutes of international calls) (total number of day calls)	Type Num Num Num Num Binary (Yes/No) Num Binary (Yes/No) Num Num Num Num Num Num Num Num Num
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number_customer_service-calls total_eve_minutes total_eve_calls churn (number of calls to customer service) (total minutes of evening calls) (total number of evening calls) (customer churn - target variable) Num Num Num Binary (Yes/No)

A data set is attached for your input to this project. A paper related to this project is also attached

Assignment:

- 1. Review the predictor variables and guess what their roles in customer churn. You are free to create new derived variables from these predictors. [Hint: You have solved similar problems in Project #3]
- 2. Divide the data into training and test set. Make sure relative proportions of true and false in the target variable are maintained in training and test set. [Hint: Use stratified random sampling: You can use data partition function of R Caret package.]
- 3. Using training data set, develop classification models using at least 3 classification techniques (1) Naïve Bayes' Classifier, 2) Any one decision tree classifier and 3) SVM classifier. You can get the classifier in R Caret package, for example, R-SVM for SVM classifier.]
- 4. Construct confusion matrices using test data set for each model. Compute a) Accuracy, b) Precision and c) Recall for each model.
- 5. Try to improve classification accuracy by choosing right set of predictor variables and model parameters and choose the best model. [Hint: Follow ROC curve, or any other if you think suitable.]

Submission procedure:

- 1. Submit R code and a report describing the process you have followed and results therein. Submit your file to Moodle course management system.
- 2. Each group should submit only one submission of the above.
- 3. Last date of submission is: **05.12.2016**, **23:55 hours**.