
Telecom Churn Prediction 2020

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Problem/Need



A telecom company puts thousands of money on a direct marketing campaign to keep its customers loyal.

- Like to know what factors cause customers to churn
- Want to spot potential churn
- Want to know what to promote on to prevent customer churn

Impact Hypothesis

- Create a classification algorithm can predict who has a high probability to churn and understand what features are contributing to churn rate the most
- Spot potential churn customer
- Know what to focus on when promoting a marketing campaign



Solution Path

1. Exploratory Data Analysis : Create diagrams to understand the dataset
2. Pick a classification metric - F1 score and Accuracy
3. Test different algorithms
4. Pick the best algorithm with the highest score on the metrics



Scoping Considerations

- **Assumption:**

People who churn share similar features

- **Measure of Success**

F1 score

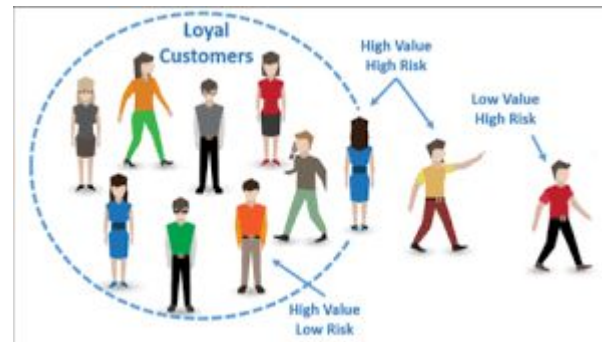
Accuracy score

- **Risk/Difficulty**

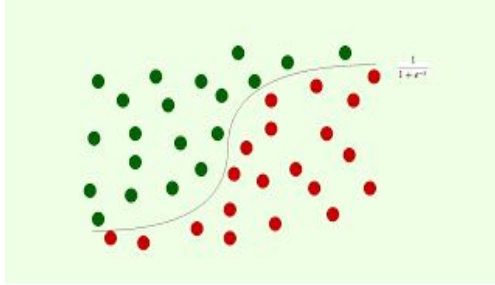
Some algorithms are hard to interpret

Data

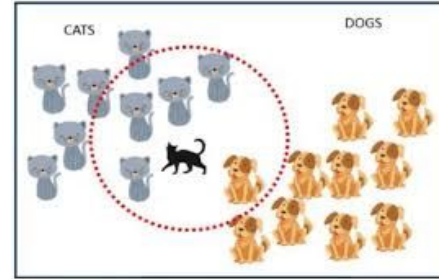
- Customer Churn Prediction 2020 from Kaggle
- Training : 4250 rows with 20 columns with one of them is the target variable
- Testing: 750 rows with same columns but not including the target variable



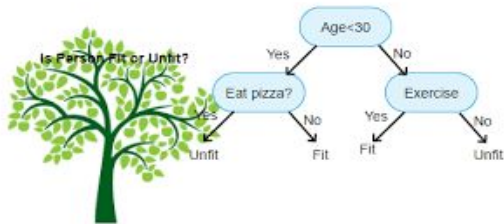
Algorithms



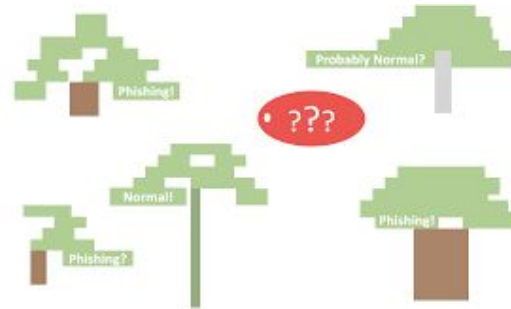
Logistic
Regression



K-Nearest
Neighbor

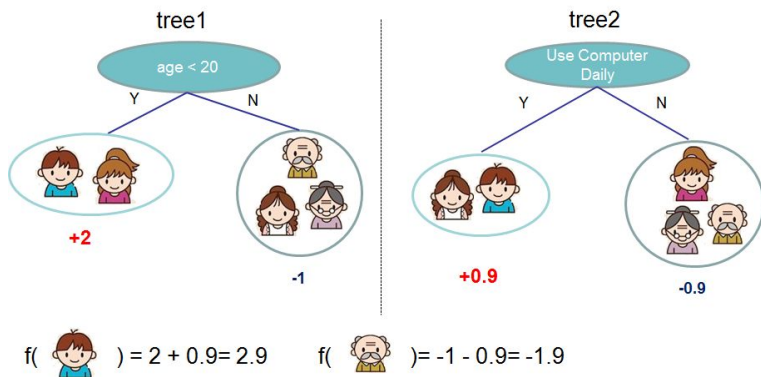


Decision
Tree



Random
Forest

Algorithms con.



Gradient Boost / XGBoost



Naive Bayes

Results on validation set

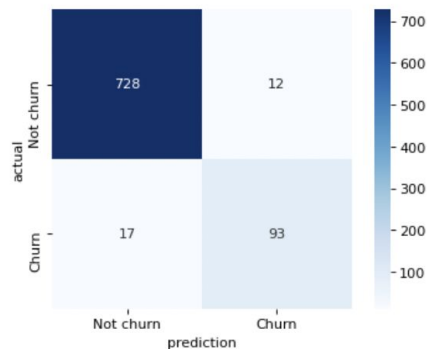
Recall score 0.6205797101449275
Precision score 0.8721121098342625
f1 score 0.7245780729893584
roc_auc score 0.9048642338797619
accuracy score 0.935686274509804

	precision	recall	f1-score	support
False	0.98	0.98	0.98	740
True	0.89	0.85	0.87	110
accuracy			0.97	850
macro avg	0.93	0.91	0.92	850
weighted avg	0.97	0.97	0.97	850

Best model is XGboost

~ 0.73 F1 score

~ 0.94 Accuracy Score



Results on actual test set

Private Score

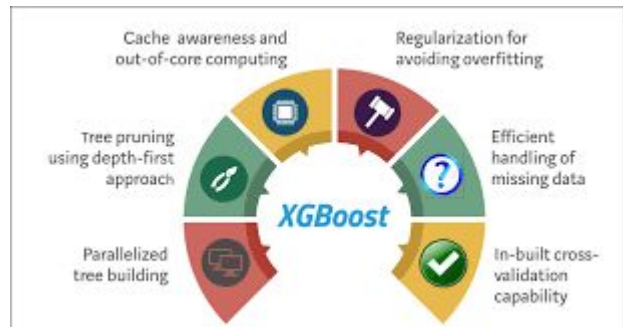
Public Score

0.93523

0.97777

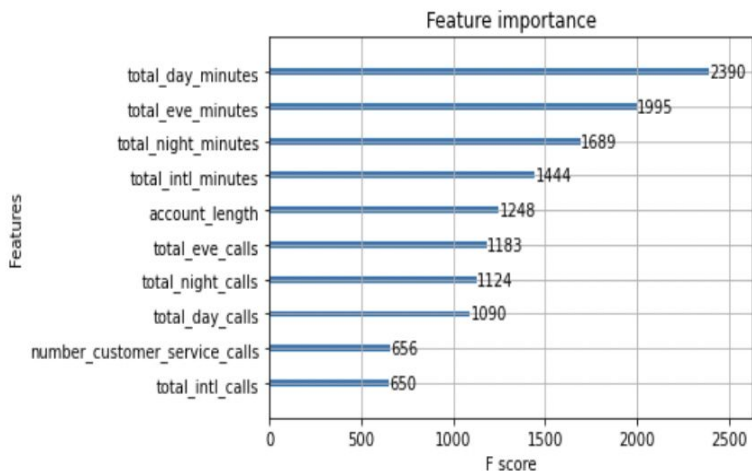
Accuracy score

~ 0.98



Feature Importances

```
<AxesSubplot:title={'center':'Feature importance'}, xlabel='F score', ylabel='Features'>
```



- Total minutes per daytime, evening, and nighttime are the top 3 important
- Account length is the 4th important

Communication

- The total minutes per day despite the time is important, so the marketing campaign can be done with giving out discounts if a customer exceeds certain minutes per month
- Account length is also important, so the telecom company can give reward to customers who have been subscribed for a certain of time



Further step

- Create marketing campaign with the slogan of giving out discount on minutes per month to attract new customers
- Reward the customers who have been subscribed for a long period of time
- Coldcall the customers who have churned to understand the reason of them churing