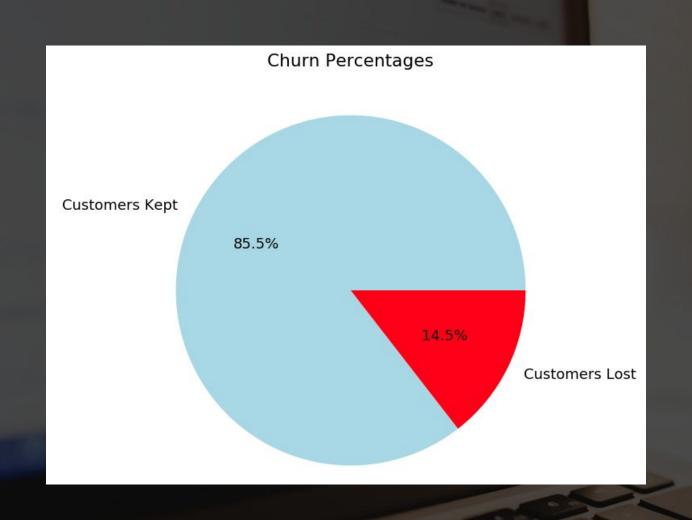
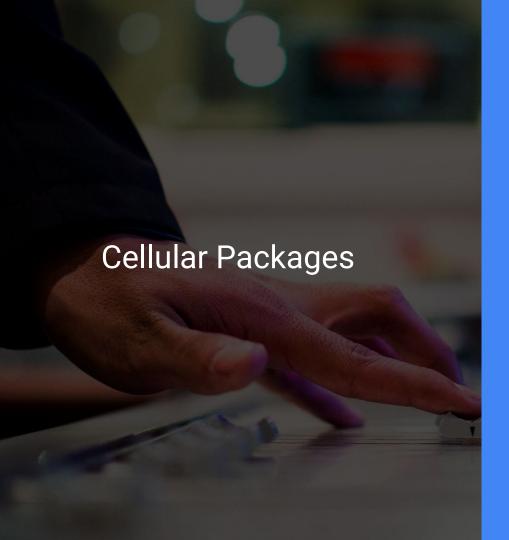
Churn at SyriaTel

Mod 3 Project Alex H. Macy



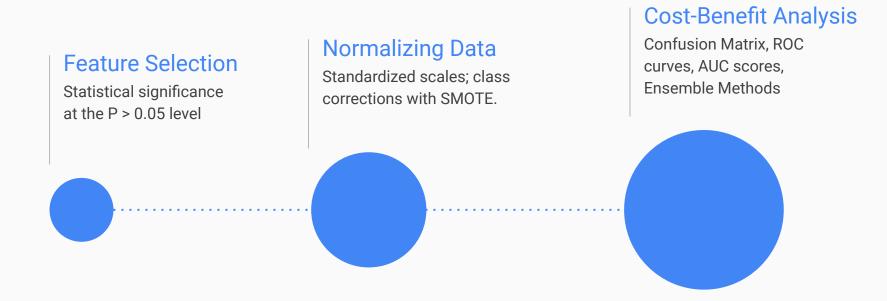


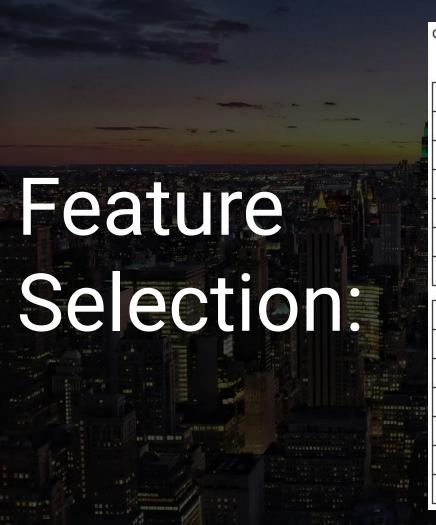


Customers across the US

- International Plans
- Voicemail Plans
 - Minutes
 - Charges

Logistic Regression





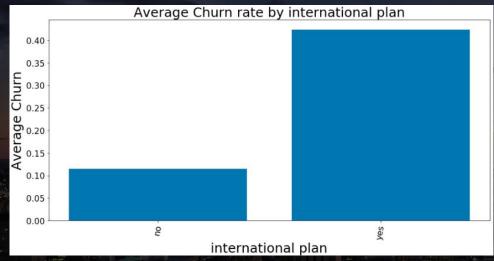
Optimization terminated successfully.

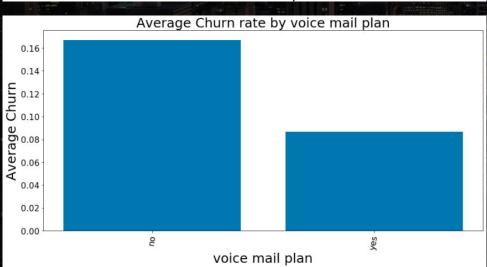
Current function value: inf

Iterations 7

Dep. Variable:	churn	No. Observations:	3333	
Model:	Logit	Df Residuals:	3327	
Method:	MLE	Df Model:	5	
Date:	Thu, 02 Apr 2020	Pseudo R-squ.:	inf	
Time:	08:57:45	Log-Likelihood:	-inf	
converged:	True	LL-Null:	0.0000	
Covariance Type:	nonrobust	LLR p-value:	1.000	

	coef	std err	z	P> z	[0.025	0.975]
const	-2.3652	0.139	-16.967	0.000	-2.638	-2.092
international plan	1.9657	0.136	14.460	0.000	1.699	2.232
voice mail plan	-1.8059	0.534	-3.381	0.001	-2.853	-0.759
number vmail messages	0.0323	0.017	1.924	0.054	-0.001	0.065
total intl calls	-0.0754	0.023	-3.214	0.001	-0.121	-0.029
customer service calls	0.4418	0.037	12.090	0.000	0.370	0.513





At first glance,

Customers with an international plan seem more likely to churn.

Let's investigate further.



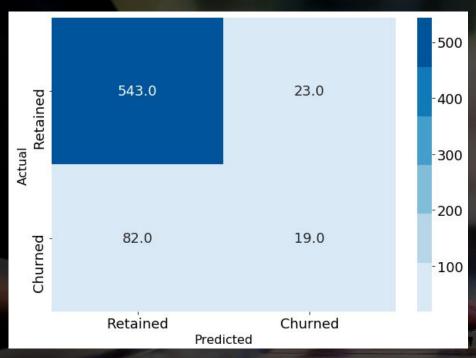
Baseline Logistic Regression Results:

		precision	recall	f1-score	support
	0	0.87	0.96	0.91	566
	1	0.45	0.19	0.27	101
accura	су			0.84	667
macro a	avg	0.66	0.57	0.59	667
eighted a	ivg	0.81	0.84	0.81	667

'Vanilla' Logistic Regression Results

84% Accuracy

Confusion Matrix

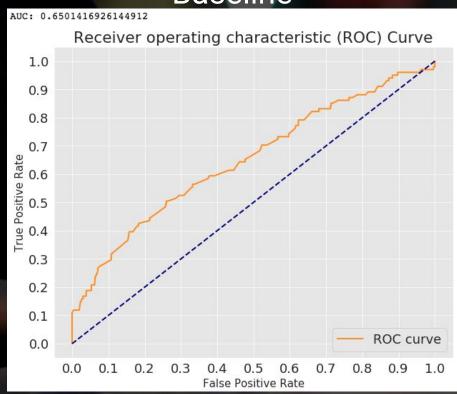


Cost-Benefit Analysis:

SyriaTel should focus on False Negatives.

Predicted 'Retained' but actually 'Churned'

Baseline



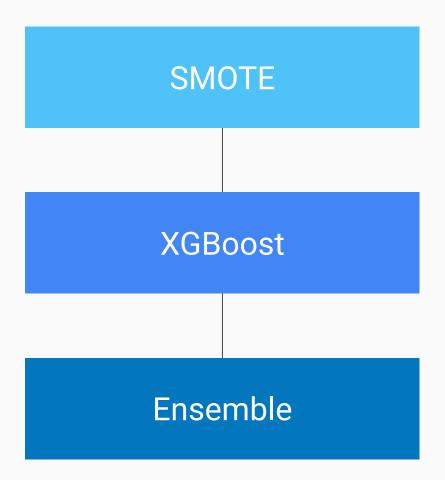
Cost-Benefit Analysis:

To gain an 80% True Positive Rate, SyriaTel would have to risk a 60% False Positive Rate.

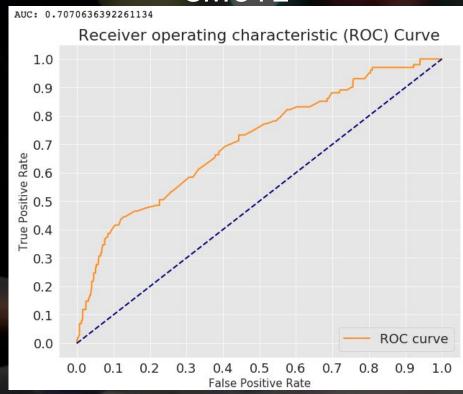
Too much risk.

Improving Performance

Shift ROC curve, and adjust AUC score



SMOTE



Cost-Benefit Analysis:

Synthetic Minority
Oversampling Technique
improved performance by 5%!

80% TPR now only risks 55% FPR

XGBoost Feature Selection:



XGBoost Performance Evaluation Before SMOTE:

XGBoost	Baseline	Classification	Report:
---------	----------	----------------	---------

	precision	recall	f1-score	support
0	0.88	0.98	0.93	566
1	0.73	0.24	0.36	101
accuracy			0.87	667
macro avg	0.80	0.61	0.64	667
weighted avg	0.86	0.87	0.84	667

Baseline LogReg Classification Report:

		precision	recall	fl-score	support
	0	0.87	0.96	0.91	566
	1	0.45	0.19	0.27	101
accur	acy			0.84	667
macro	avg	0.66	0.57	0.59	667
weighted	avg	0.81	0.84	0.81	667

XGBoost Performance Evaluation After SMOTE:

XGBoost	SMOTE	Classification	Report:

	precision	recall	fl-score	support
0	0.90	0.92	0.91	566
1	0.47	0.42	0.44	101
accuracy			0.84	667
macro avg	0.68	0.67	0.67	667
weighted avg	0.83	0.84	0.84	667

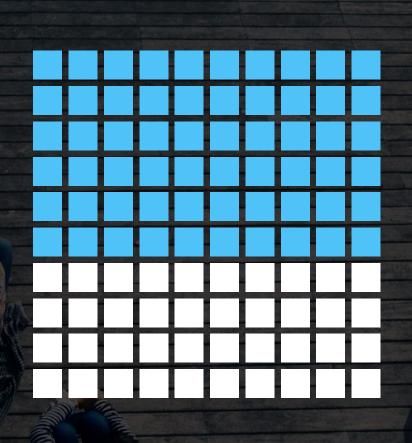
XGBoost Baseline Classification Report:

		precision	recall	f1-score	support
	0	0.88	0.98	0.93	566
	1	0.73	0.24	0.36	101
accui	racy			0.87	667
macro	avg	0.80	0.61	0.64	667
weighted	avg	0.86	0.87	0.84	667

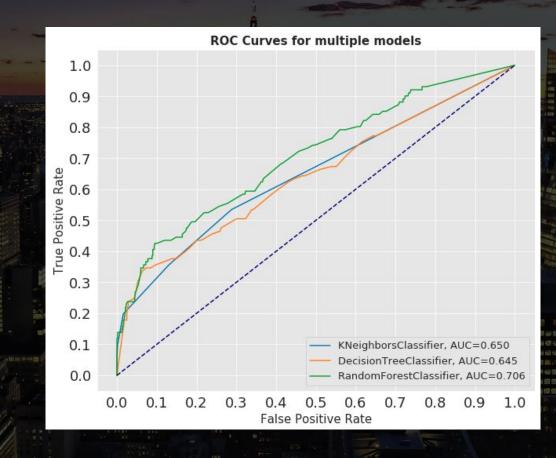
XGBoost

With Synthetic Minority Oversampling was able to reach 84% accuracy on our data

Ensemble methods may do better



Baseline Ensemble Method Performance Example



Recommendations:

Model churn data across multiple models using

- Different forms of Feature
 Selection
 - KBest
 - Recursive FeatureElimination
 - Principal Component Analysis
- Cross Validation techniques

