

Napster 2.0 Churn

Predicting customer churn rate for the next month

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The problem

Company

Napster 2.0 is a music streaming service.

Context

Subscribers pay **\$10** for access to the service monthly.

Problem statement

Customers have an average lifetime of only **1.75 months**.

Executive Summary

Revenue impact

Offering a 30% discount to potential churners can **improve our revenue by 7%** under 30% acceptance rate.

Scope of Offering

We should offer discount to those with **more than 60% probability of churning** as predicted by our model

Solution

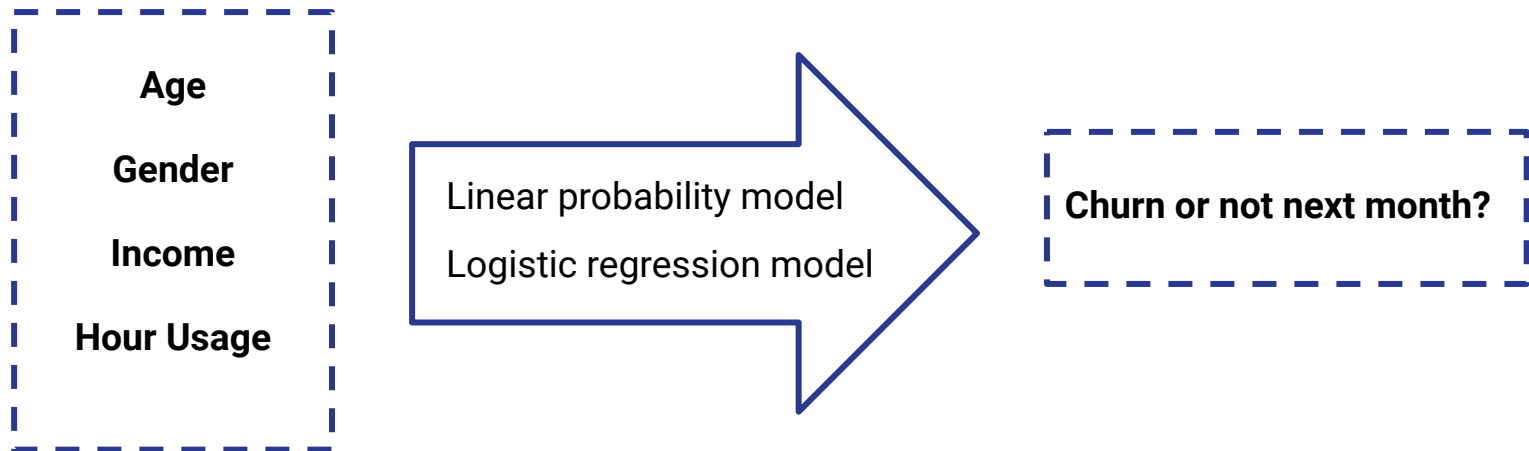
Offer discount to churners

According to our models and projected revenues, we can see that offering discounts to potential churners will extend customer lifetime value and increase our projected revenue.

Implementation

Solution

Predictive modeling to identify those who are likely to churn



Linear probability model

OLS Regression Results

```
=====
Dep. Variable:          churn_TF      R-squared:                0.122
Model:                  OLS           Adj. R-squared:          0.122
Method:                 Least Squares F-statistic:             244.1
Date:                  Tue, 17 Nov 2020 Prob (F-statistic):      1.46e-196
Time:                  10:09:22       Log-Likelihood:         -4546.0
No. Observations:      7000          AIC:                   9102.
Df Residuals:          6995          BIC:                   9136.
Df Model:               4
Covariance Type:       nonrobust
=====
```

```
=====
              coef      std err          t      P>|t|      [ 0.025      0.975 ]
-----
age           0.1309      0.006     23.618      0.000      0.120      0.142
income       -0.0903      0.006    -16.391      0.000     -0.101     -0.080
hours       -0.0636      0.006    -11.531      0.000     -0.074     -0.053
female       0.0590      0.011      5.322      0.000      0.037      0.081
const       0.5465      0.008     70.410      0.000      0.531      0.562
=====
```

Logistic regression model

Logit Regression Results

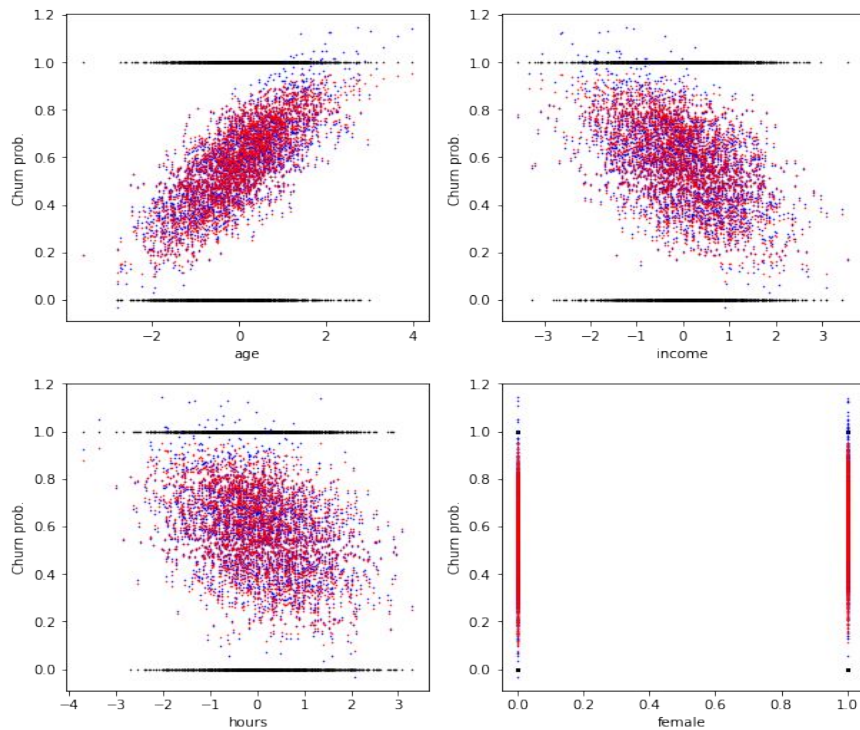
```
=====
Dep. Variable:          churn_TF      No. Observations:          7000
Model:                  Logit         Df Residuals:              6995
Method:                  MLE          Df Model:                  4
Date:                   Tue, 17 Nov 2020  Pseudo R-squ.:          0.09580
Time:                   08:34:50         Log-Likelihood:           -4317.9
converged:              True           LL-Null:                  -4775.4
Covariance Type:        nonrobust      LLR p-value:              9.545e-197
=====
```

```
=====
              coef      std err          z      P>|z|      [0.025      0.975]
-----
age           0.6109      0.028      21.887      0.000      0.556      0.666
income        -0.4219      0.027     -15.818      0.000     -0.474     -0.370
hours         -0.2967      0.026     -11.315      0.000     -0.348     -0.245
female         0.2755      0.052       5.321      0.000      0.174      0.377
const         0.2113      0.036       5.876      0.000      0.141      0.282
=====
```

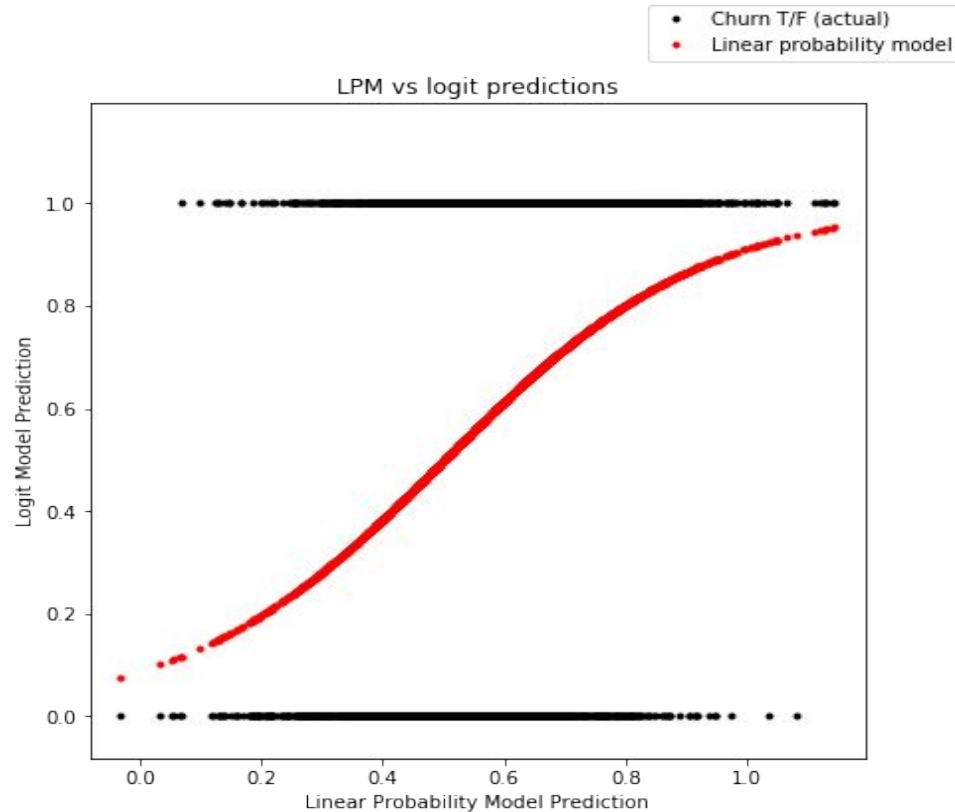

Feature importance

Single factor relationship with churn

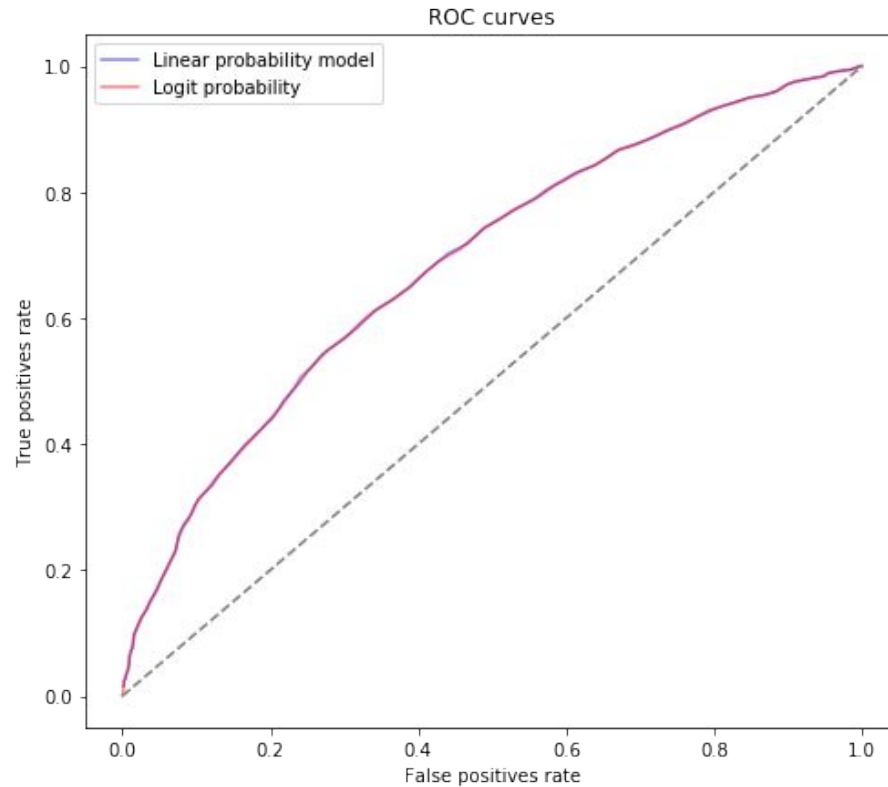
- Churn T/F (actual)
- Linear probability model
- Logit probability



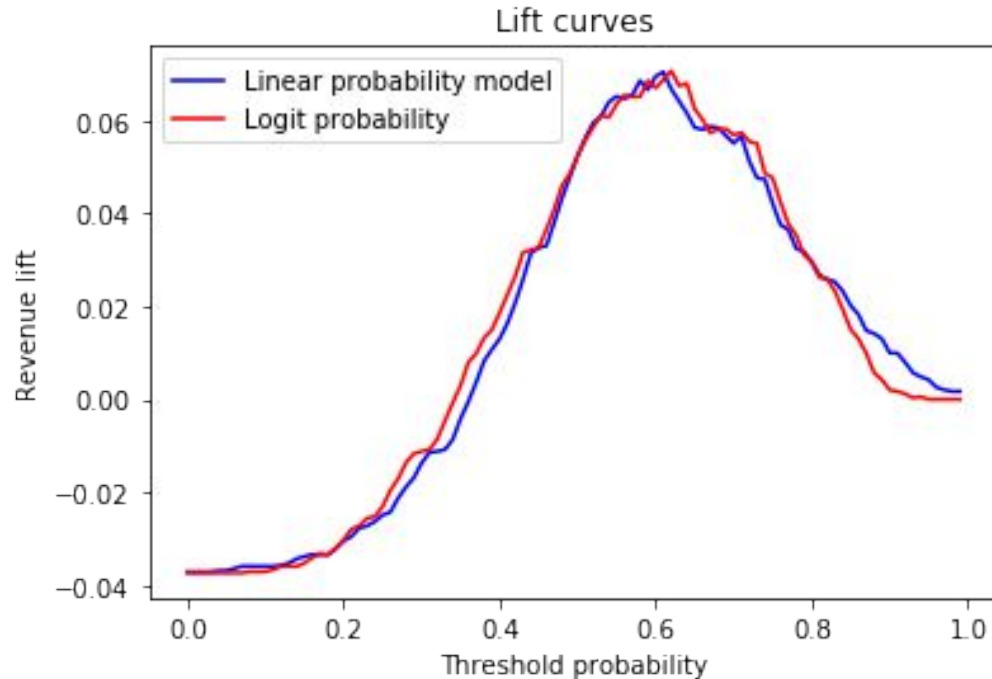
Model consistency



ROC curve



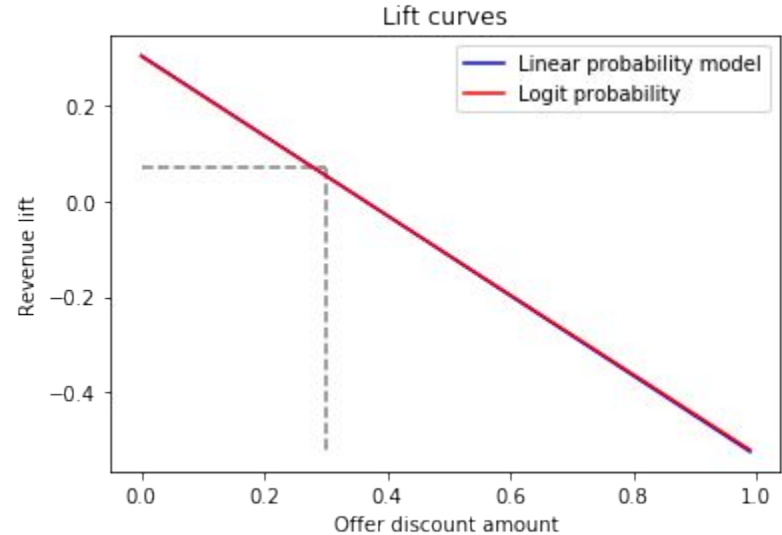
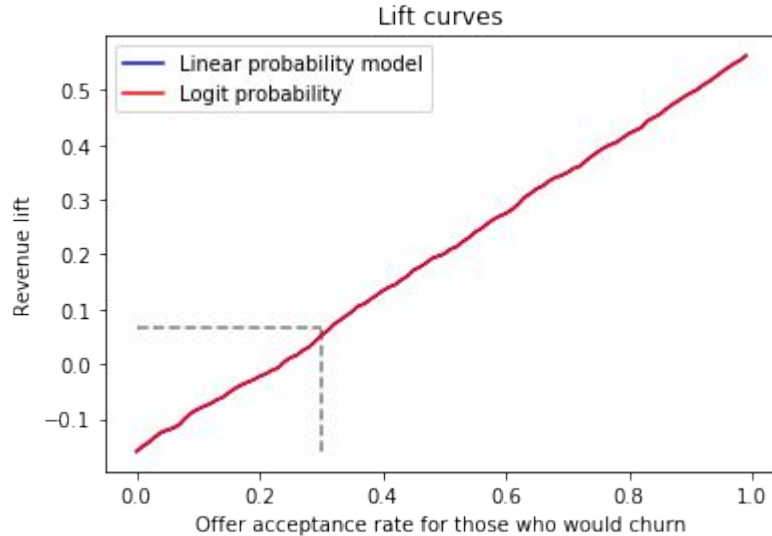
Impact of probability threshold on revenue



Under assumptions:

1. Discount amount 30%
2. Acceptance rate 30%

Sensitivity of our assumptions



Under assumptions: probability threshold 50%

Other issues

- 1. How much should the discount be?**

It depends on the relationship between discount rate and acceptance rate.

- 2. What are channels to offer discount to potential churners?**

Email could have a higher acceptance rate than phone call

- 3. Offering discount is expensive, how to cut more cost?**

Call or send emails checking customer satisfaction before making the offer

Thank you!

Any question?

