

Churn Ultimatum

> On a mission to uncover hidden truths about customer retention

#### **Table of Contents**



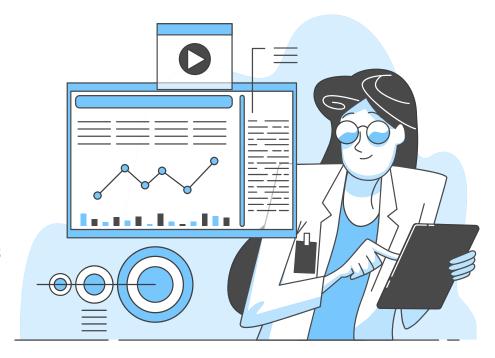
#### Business Case & Data Overview



**Key Insights** 



**Conclusions & Next Steps** 





## **Understanding Churn**





#### **Identify Precisely**

Focus on customers who will most likely churn

#### **Wasted Resources**

Ensure effective allocation of resources





## **Model Performance**

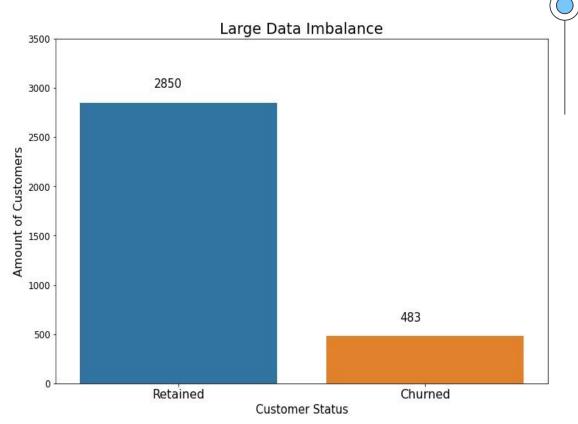
	Predicted: No-Churn	Predicted: Will Churn
Actual: No-Churn	Predicted not to churn correctly	Predicted to churn but actually not going to churn
Actual: Will Churn	Customer will churn and model did not predict it.	Predicted to churn correctly



# **Large Data Imbalance**

Much **fewer churned** data points

Can be **difficult to**predict a minority class





# **Dummy Model**



**128** times predicted customer would churn but did not.....

Predicted:

Predicted:

No-Churn

Will Churn

... or **128** times resources wasted

Actual:

**No-Churn** 

Predicted not to churn correctly

**727** 

Predicted to churn incorrectly

128

0.13
Precision score

Actual:

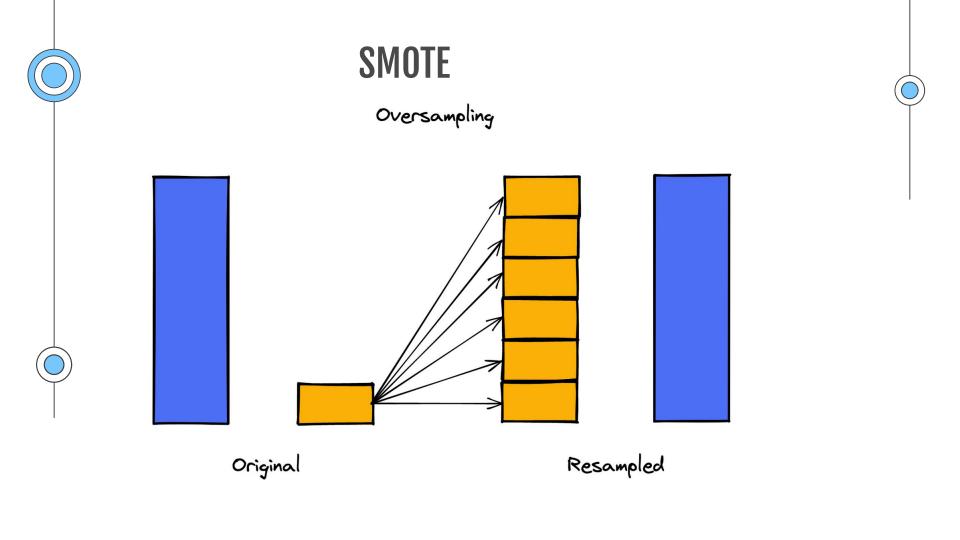
Will Churn

Predicted to not churn incorrectly

123

Predicted to churn correctly

22





#### Final Model: test results



**0** times predicted customer would churn... Model may be **overfit** 

Predicted:

Predicted:

No-Churn

Will Churn

Final model does not predict all churn

Actual:

**No-Churn** 

Predicted not to churn correctly

855

Predicted to churn incorrectly

0

1.0 precision (.91 cross-val)

Actual:

Will Churn

Predicted not to churn incorrectly

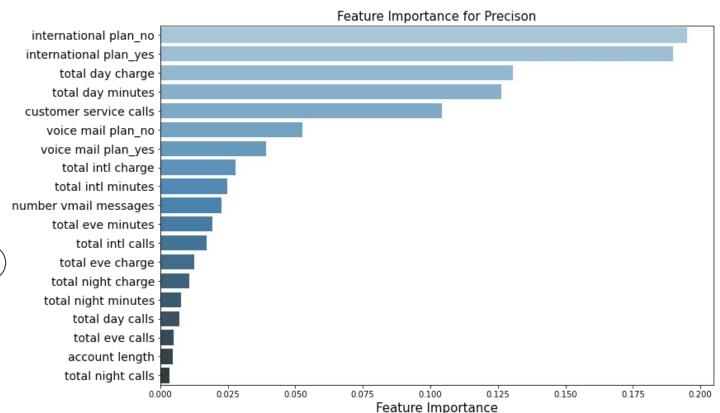
123

Predicted to churn correctly

21



### Final Features



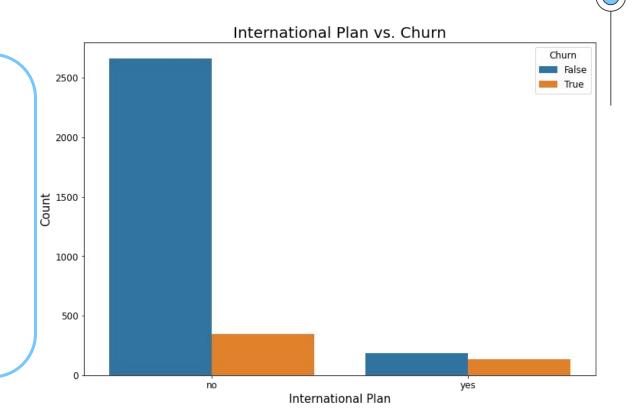


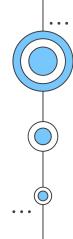
#### Final Features

Model shows highest importance for your **international plan** 

Likely an area you will want to focus on

May want to consider an **indirect approach** 





# Conclusions

Precision allows you to be confident with resource allocation

 Model is **precise** in targeting ACTUAL churn and not targeting FALSE churn. But still misses a lot of churn

 Model is more precise when it comes to international plan indicating this could be an area of concern.

#### Feature Engineering

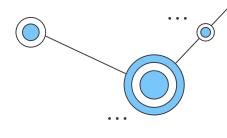
Experiment by adding, removing, merging or transforming features

# **Next Steps**

#### Adjusting Model for

#### recall

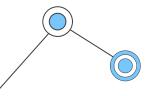
Tweak the model to miss less positive cases by trying different algorithms and parameters



#### Analyze location features

Certain locations could be more competitive or provide worse coverage





# Thanks!

# Do you have any questions?

For additional info, contact Max Ross on GitHub (https://github.com/ImMaxRoss)

