



# DECODING **CHURN**

A Data-Driven Strategy for Customer  
Retention

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# BUSINESS UNDERSTANDING: THE PROBLEM

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## REVENUE LOSS

SyriaTel is losing revenue due to customer churn. This is a significant challenge because **acquiring new customers is more expensive than retaining existing ones**.

Current retention strategies are reactive, often addressing the issue only after the customer has already decided to leave.

## THE NEED

A data-driven approach is needed to identify customers at **high risk of churning** before they leave.

This will allow the marketing and retention teams to apply targeted strategies more effectively, optimizing the budget and saving revenue.


# PROJECT OBJECTIVES


## PRIMARY GOAL


### **Maximize Recall.**

Our priority is to catch as many potential churners as possible. We prioritize minimizing "False Negatives" (missed churners) over precision, as losing a customer is more costly than a retention offer.

## SECONDARY GOALS

 **Identify Key Features** Uncover drivers like pricing and service quality to inform retention strategy.

 **Compare Models** Evaluate algorithms to balance the Precision-Recall tradeoff effectively.

 **Assess Performance** Assess model performance using business-relevant evaluation metrics.

# DATA OVERVIEW: PROFILE & BALANCE

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3,333

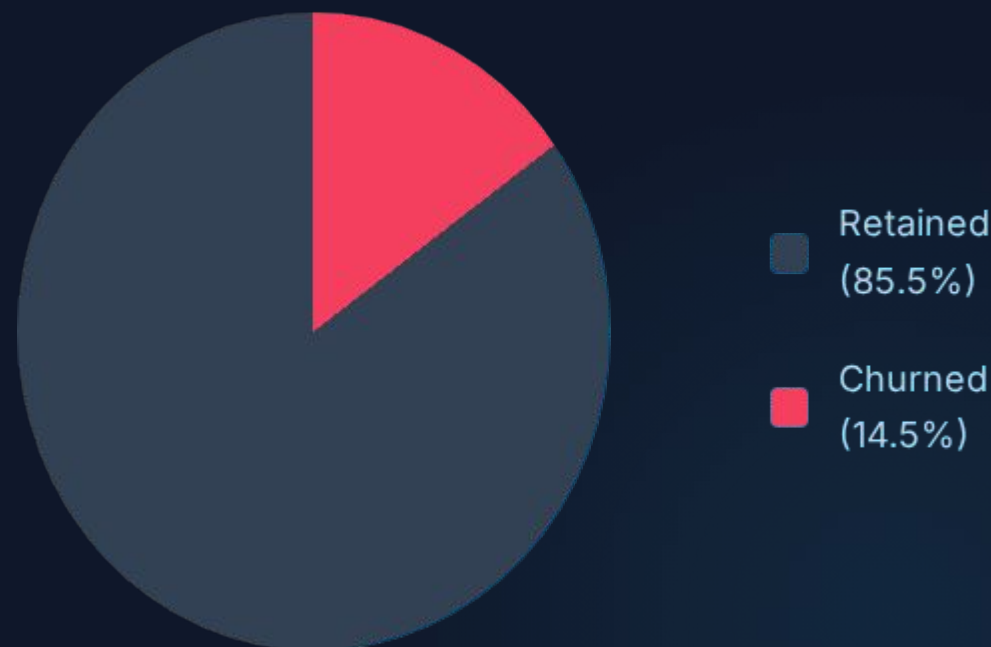
TOTAL CUSTOMERS

Complete historical usage records

21

FEATURES RECORDED

Including call minutes, plans, & service calls



**The Churn Reality:** Only 14.5% of customers churned. This "Class Imbalance" makes accuracy a misleading metric for success.

# BEHAVIORAL FEATURE ENGINEERING

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To improve predictive power, we moved beyond raw data to create advanced behavioral features:



**Total Minutes** This represents the customer's overall reliance on the network regardless of time.



**Total Charges** This serves as a proxy for price sensitivity—higher spenders are often the first to churn if value drops.



**Peak Period** Identified the specific time of day (Day, Evening, Night, or Intl) where the customer has the highest usage intensity.



**Day-Night Ratio** Captured the balance between peak and off-peak usage to detect unusual lifestyle patterns or outliers.



**Intl Usage Flag** A binary indicator for high-intensity international users, helping segment customers with specific global connectivity needs.

# METHODOLOGY: THE "ANTI-LEAKAGE" STRATEGY

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## PREPROCESSING

Applied **StandardScaler** and **OneHotEncoder** to ensure the model treats all numerical magnitudes and categories fairly without bias.



## SMOTE STRATEGY

Used **Synthetic Minority Over-sampling Technique** to "teach" the model what a churner looks like by creating synthetic data points.



## INTEGRITY CHECK

Crucially, SMOTE was applied **only to the training data**. The test set remains pure to ensure our performance results are realistic.

# MODEL COMPARISON RESULTS

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## LOGISTIC REGRESSION

**The Baseline.** Provided a good starting point but struggled to capture complex, non-linear customer behaviors.



## DECISION TREE

**The Learner.** Better at capturing patterns but highly prone to "overfitting"—memorizing the data rather than learning from it.



## RANDOM FOREST

**The Champion.** An Ensemble Model that combines multiple perspectives to provide the most stable, reliable, and high-recall predictions.

# PERFORMANCE RESULTS

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## THE OUTCOME

**Model: Final Hypertuned Random Forest (Threshold: 0.65)**

Our optimized model successfully identifies a significant majority of customers who are about to leave.

- ✓ **Recall Score: ~66%**  
(Captures approx. 3 out of 5 churners)
- ✓ **F1 Score: ~74%**  
(Balanced precision and recall)



### RECALL MAXIMIZATION

By prioritizing Recall, we accept some "False Alarms" to ensure we don't miss the expensive loss of a customer.






# TOP DRIVERS OF CHURN (FEATURE IMPORTANCE)



**Insight:** Customers with high bills who have called support 3+ times are in the "Danger Zone".

# RECOMMENDATIONS FOR RETENTION

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-  **High-Usage Discounts** Proactively offer loyalty discounts to the "High Charge" segment identified by the model. Don't wait for them to complain.
-  **The "3-Call" Rule** Flag any customer who makes their 3rd service call for immediate follow-up by a senior retention specialist.
-  **International Plan Audit** Review the value proposition of the international plan. It is currently a "churn magnet," suggesting pricing or connection quality issues.

# CONCLUSION & NEXT STEPS

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## ADDRESS KEY DRIVERS OF CHURN

A/B test retention offers (e.g., 10% discount vs. free data) to minimize churn.



## ACTION PLAN

Deploy the model to flag customers with >3 service calls and offer proactive "Usage Discounts" to high spenders.

"Retention starts with Detection"



# THANK YOU

A detailed analysis can be found on my

github: <https://github.com/NjorogeWinnie/Telecom-churn-analysis>

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