

JAZA TELECOM COMPANY CUSTOMER CHURN PREDICTION MODEL

NAME: PETER KINYANJUI

INSTITUTION: MORINGA SCHOOL

INSTRUCTOR: FAITH ROTICH

QUOTE:
GOOD CUSTOMER SERVICE COSTS
LESS THAN CUSTOMER CHURN

- ANONYMOUS



OVERVIEW

1. Customer Centrality:

-Success in telecom hinges on customer satisfaction and loyalty.

2. Churn Signals:

-If customers switch due to fees, better offers, or poor service, it reflects telecom performance.

3. Performance Indicator:

-Customer churn indirectly gauges telecom effectiveness and competitiveness.

4. Critical Metric:

-Monitoring churn is vital for assessing telecom companies' market standing and operational efficiency.

BUSINESS PROBLEM

1. Customer Retention Focus:

- Jaza Telecom historically prioritized getting new customers over keeping existing ones.

2. Cost Dynamics:

- Acquiring new customers can cost up to five times more than retaining current ones.

3. Profit Potential:

- Research shows a 5% increase in retention can boost profits by 25% to 95%.

4. Churn Impact:

- Failure to identify and address customer churn signals can harm the company's long-term performance.

5. Data Insights:

- Customer churn data offers valuable insights and fuels predictive models for proactive actions.



DATA UNDERSTANDING

1. Source of Information:

- Data comes from <https://www.kaggle.com/datasets/becksddf/churn-in-telecoms-dataset> , providing details on customer interactions.

2. Dataset Overview:

- Contains 3333 entries and 21 columns with information like account length, usage metrics, and customer service calls.

3. Data Types:

- Various data types include object (e.g., state, phone number), int64, float64, and bool (for churn).

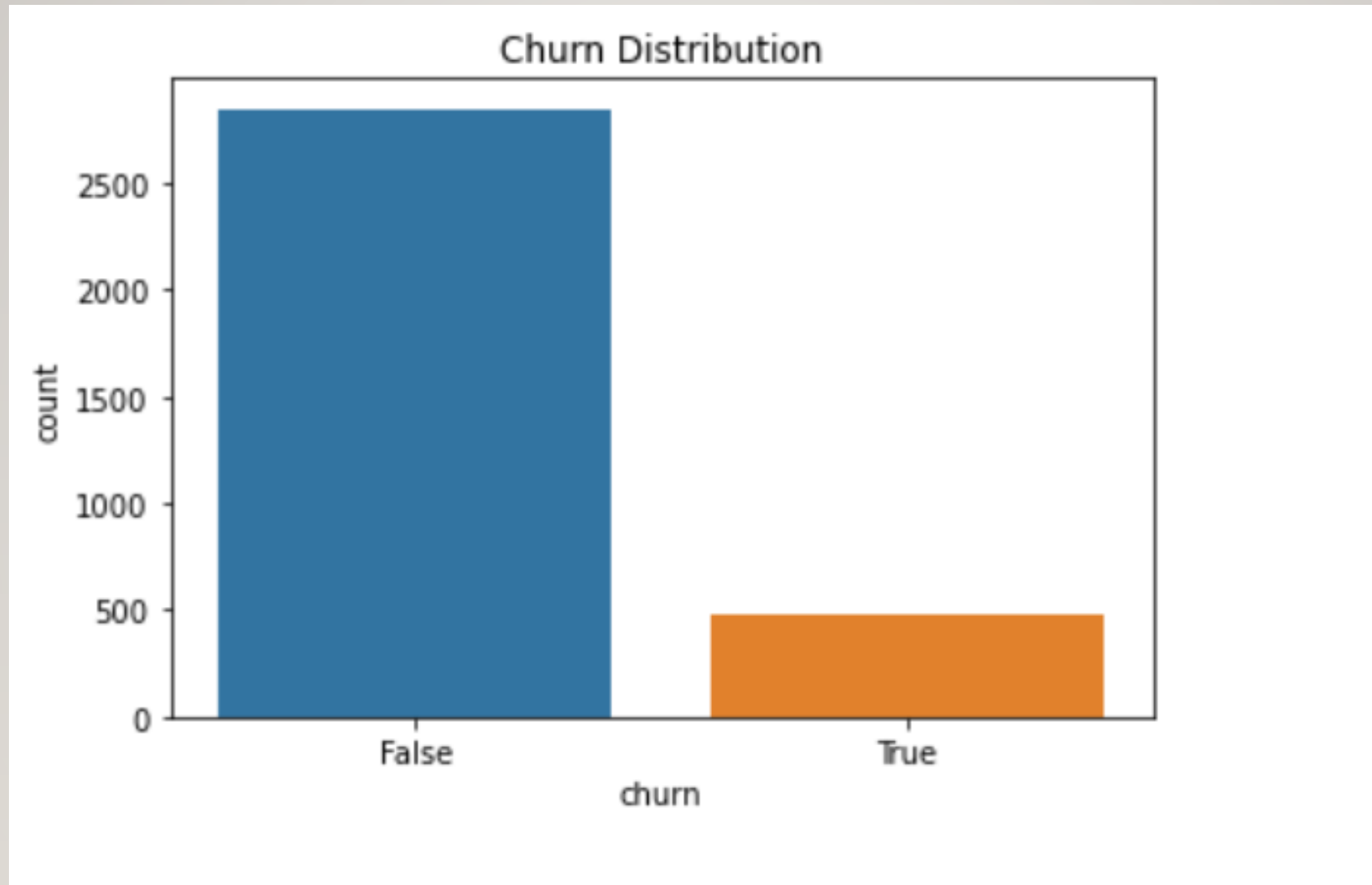
CONTINUATION...

4. Data Quality:

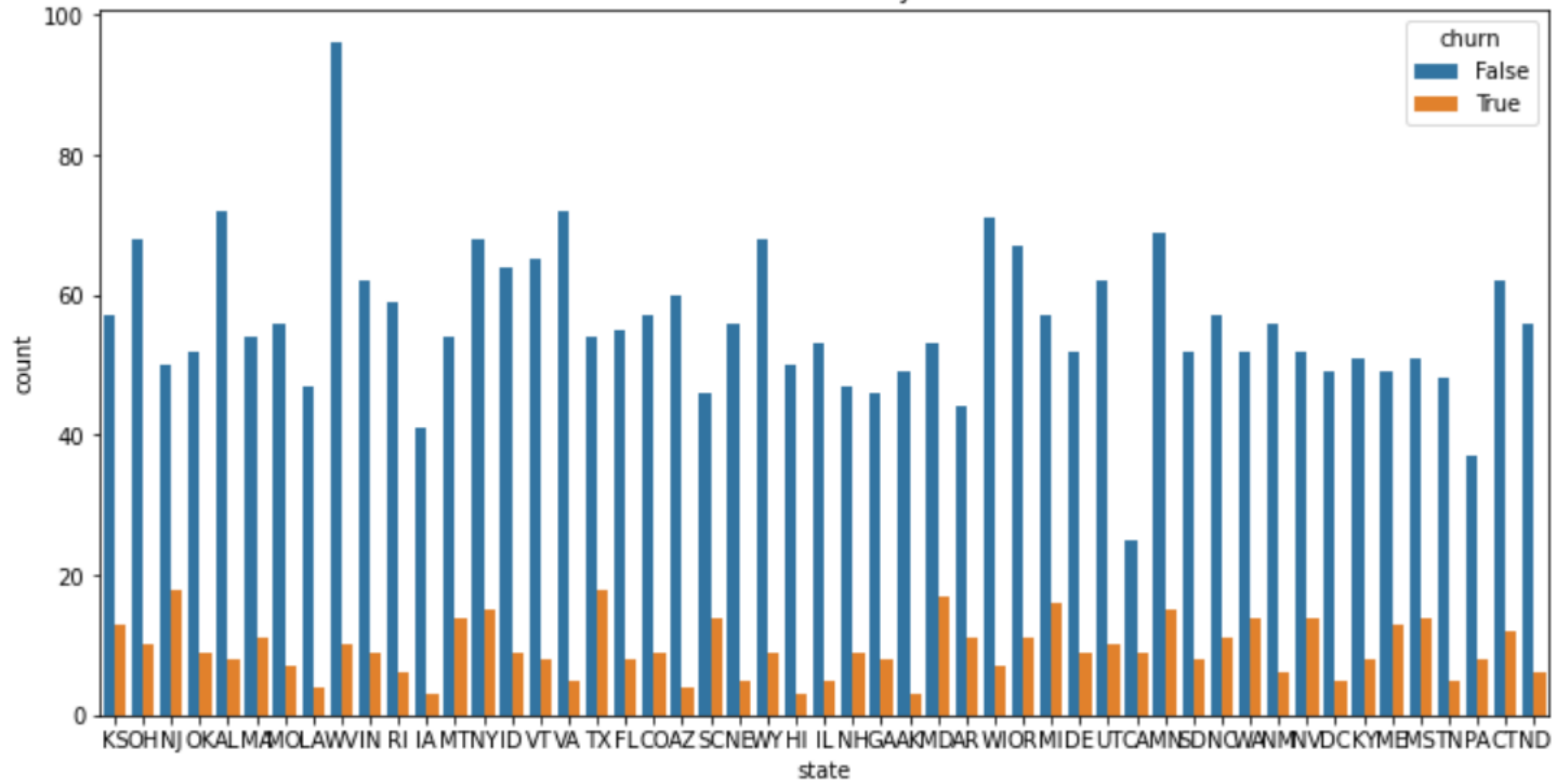
- No missing values in the dataset.
- Categorical variables may need encoding for model training.

5. Target Variable:

- 'Churn' column, a boolean, indicates if a customer has discontinued services.
- A rich dataset offering insights into customer behavior, serving as a foundation for predicting and addressing churn.



state Distribution by Churn



MODELING

1. Simple model - Logistic regression
2. Complex Model- Random forest
3. Hyperparameter tuned for Random forest model

EVALUATION- LINEAR REGRESSION MODEL

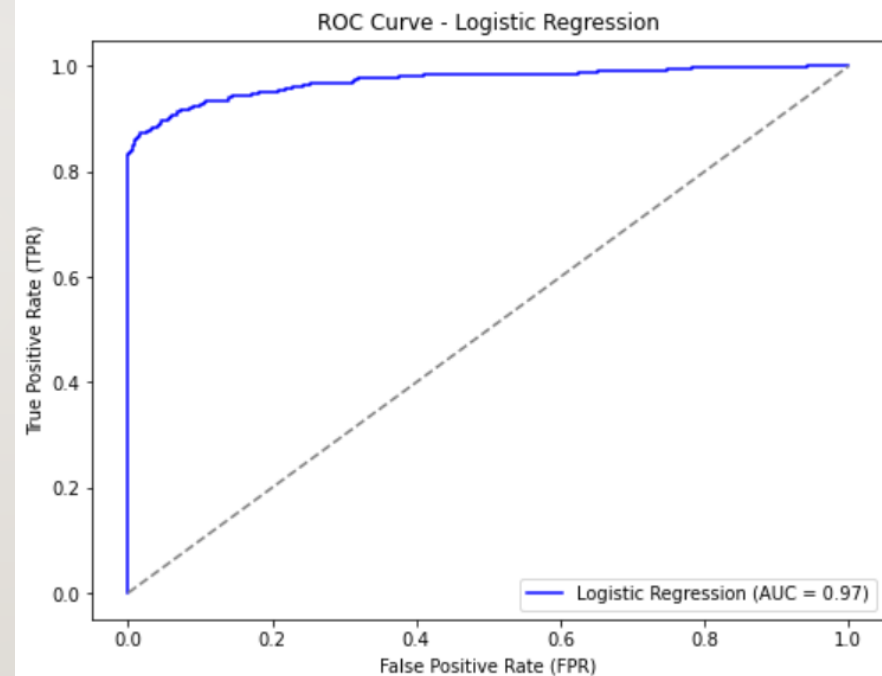
Logistic Regression Accuracy: 0.92

Confusion Matrix:

```
[[560  25]
 [ 63 492]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.90	0.96	0.93	585
1	0.95	0.89	0.92	555
accuracy			0.92	1140
macro avg	0.93	0.92	0.92	1140
weighted avg	0.92	0.92	0.92	1140



1.Accuracy:

Achieved a commendable 92% accuracy in predicting customer churn.

2.Confusion Matrix:

Balanced performance with 25 false positives and 63 false negatives.

3.Precision-Recall:

Precision of 95% for non-churn and 90% for churn.

Balanced recall of 89% for churn and 96% for non-churn.



EVALUATION- RANDOM FOREST MODEL

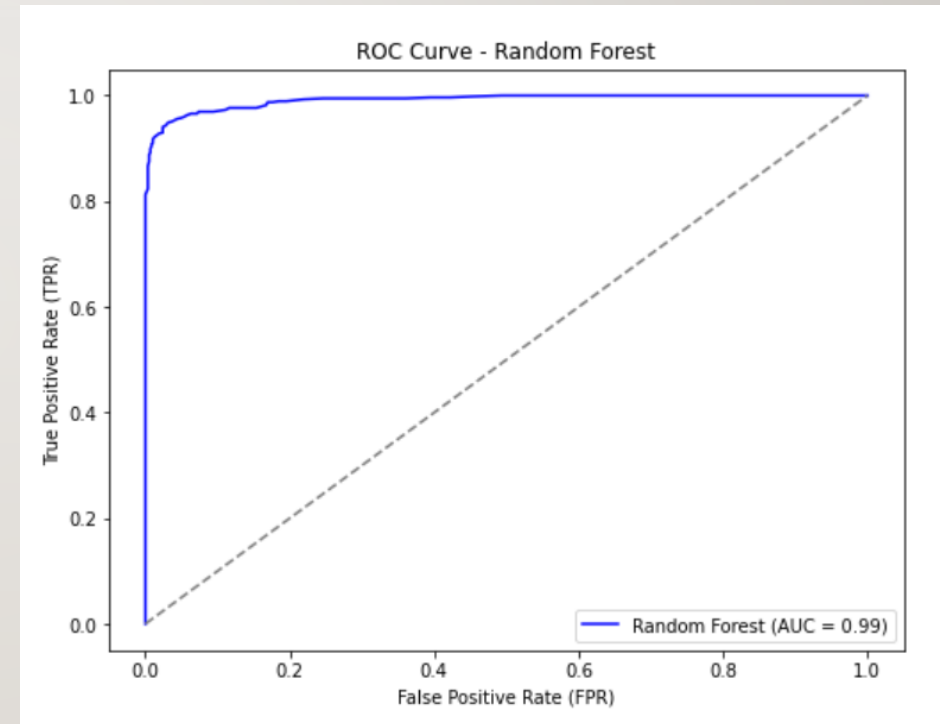
Random Forest Accuracy: 0.96

Confusion Matrix:

```
[[565  20]
 [ 28 527]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.95	0.97	0.96	585
1	0.96	0.95	0.96	555
accuracy			0.96	1140
macro avg	0.96	0.96	0.96	1140
weighted avg	0.96	0.96	0.96	1140



1.Accuracy Boost:

Demonstrated superior performance with a higher accuracy of 96%.

2.Confusion Matrix Improvement:

Significantly reduced false positives (20) and false negatives (28).

3.Precision-Recall Enhancement:

Maintained high precision and recall for both churn and non-churn classes.



EVALUATION- HYPERPARAMETER TUNED FOR RANDOM FOREST MODEL

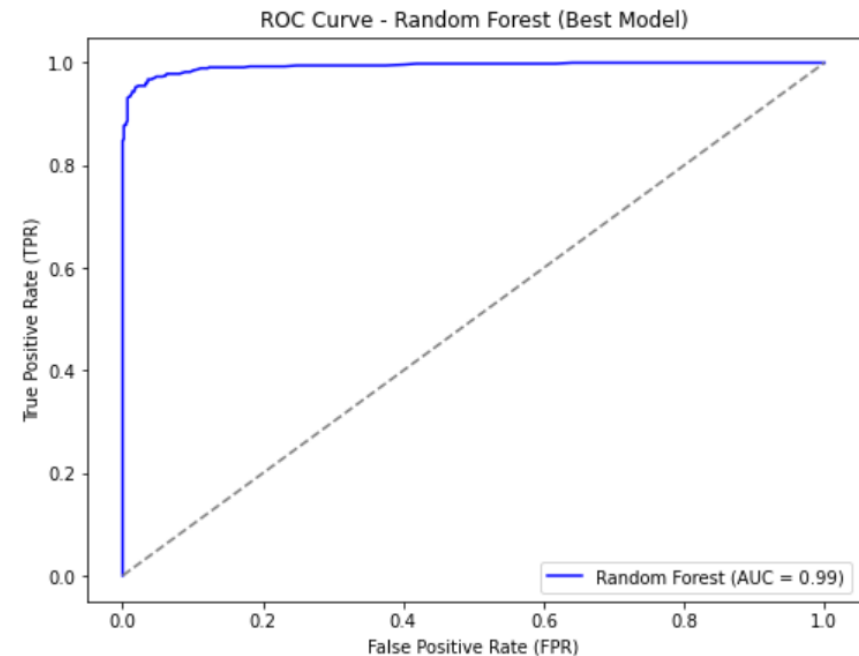
Hyperparameter-Tuned Random Forest Accuracy: 0.96

Confusion Matrix:

```
[[566 19]
 [ 22 533]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.96	0.97	0.97	585
1	0.97	0.96	0.96	555
accuracy			0.96	1140
macro avg	0.96	0.96	0.96	1140
weighted avg	0.96	0.96	0.96	1140



1.Optimized Accuracy:

Maintained high accuracy of 96%, matching the untuned model.

2.Further Confusion Matrix Refinement:

Continued reduction in false positives (19) and false negatives (22).

3.Precision-Recall Consistency:

Precision and recall metrics remained consistently high.

4.Visual Validation:

Confusion matrix heatmap and ROC curve illustrated the continued effectiveness of the tuned model.



RECOMMENDATION

- By incorporating these recommendations, Jaza Telecom can effectively integrate predictive modeling into its operational and strategic decision-making processes, ultimately improving customer retention and overall business performance.

1. Deployment Strategy:

Consider deploying the hyperparameter-tuned Random Forest model for its superior accuracy and optimized performance.

2. Continuous Monitoring:

Implement a system for ongoing model monitoring to ensure sustained effectiveness in predicting customer churn.



CONTINUATION...

3. Strategic Insights:

Leverage insights from model evaluations to inform strategic decisions related to customer retention and satisfaction.

4. Collaboration with Teams:

Collaborate with marketing and customer service teams to implement targeted strategies based on model predictions.

5. Training and Communication:

Conduct training sessions to familiarize relevant teams with the models, their predictions, and the associated business implications.



NEXT STEP

1. Model Deployment:

Roll out the selected model (hyperparameter-tuned Random Forest) into our production environment for real-time predictions.

2. Integration with Systems:

Integrate the model seamlessly with existing operational systems to ensure a smooth flow of information and decision-making.

3. User Interface Development:

Develop a user-friendly interface for teams to interact with the model predictions, making insights easily accessible and actionable.

4. Automation of Actions:

Implement automated processes triggered by model predictions to streamline and optimize customer retention strategies.

5. KPI Monitoring:

Set up key performance indicators (KPIs) to monitor the effectiveness of the model-driven strategies and regularly evaluate against predefined benchmarks.



Thank you

