

- 3. Fit a Decision tree Model
- 4. Fit a Decision tree Tuned Model (Improve the previous model)
- 5. Compare the models
- 6. Determine feature importanceestablished universes.

Visualization

We utilized a variety of visual tools to present our findings clearly:

Confusion matrix for a better obeservation of TP FP FN TN Bar Chat to show best model Bar plot to display feature importance

Recommendations

After training and analysis we come out with some insights that could help predict and potentially reduce churn in the company business.

- 1. Best Model Recommendation would be Tuned Decision tree since it has the highest ROC AUC score compared to Logistic Regression.
- 2. Customers who spend more time per day tend to churn the most, that may due to the service charging fees. Therefore I would recommend to propose affordable service bundle.
- 3. Another factor which is related to churn is customer service calls. Either the customer is not satify after calling, or he cannot find people to take his case in consideration. Then I would recommend to improve customer service experience.

Code Quality

All code was written in Python using industry-standard libraries such as Pandas, Matplotlib, and skeatlearn . Key characteristics of the codebase include:

- Modularity: Functions are separated for loading, cleaning, analyzing, and visualizing data to promote reusability.
- Documentation: Inline comments and docstrings are used to clarify purpose and logic.
- Efficiency: Vectorized operations and optimized queries reduce runtime for large datasets.
- Reproducibility: Code is organized in a Jupyter Notebook, allowing anyone to rerun the analysis with minimal setup.
- Version Control: Git was used to track changes and manage collaboration.

Repo Structure

Data/churn_dataset.csv Partials/*.py [some modules] PDFs/Github.pdf PDFs/Notebook.pdf PDFs/Presentation.pdf .gitignore LICENSE Main.ipynb README.md