# Customer Churn Prediction Using Machine Learning

Telco Customer Churn Analysis with Logistic Regression, Random Forest, and XGBoost

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## **Problem Statement**

#### **Business Question:**

- Identify factors influencing customer churn.
- Predict which customers are likely to leave.

#### Why It's Important:

- Losing customers increases acquisition costs.
- Retention improves revenue and customer lifetime value.

### **Dataset Overview**

**Rows:** 7,043. **Columns:** 21. **Source:** Kaggle – Telco Customer Churn dataset

#### **Key Feature Groups:**

- Demographics (gender, senior citizen, dependents)
- Service details (Internet type, security, tech support)
- Account info (contract, payment method, paperless billing)
- Financials (monthly & total charges)
- Target: Churn (Yes/No)

	customerID	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	MultipleLines	InternetService	OnlineSecurity	 DeviceProtection	TechSupport	Sti
0	7590- VHVEG	Female	0	Yes	No	1	No	No phone service	DSL	No	 No	No	
1	5575- GNVDE	Male	0	No	No	34	Yes	No	DSL	Yes	 Yes	No	
2	3668- QPYBK	Male	0	No	No	2	Yes	No	DSL	Yes	 No	No	
3	7795- CFOCW	Male	0	No	No	45	No	No phone service	DSL	Yes	 Yes	Yes	
4	9237- HQITU	Female	0	No	No	2	Yes	No	Fiber optic	No	 No	No	

## **Data Cleaning & Preparation**

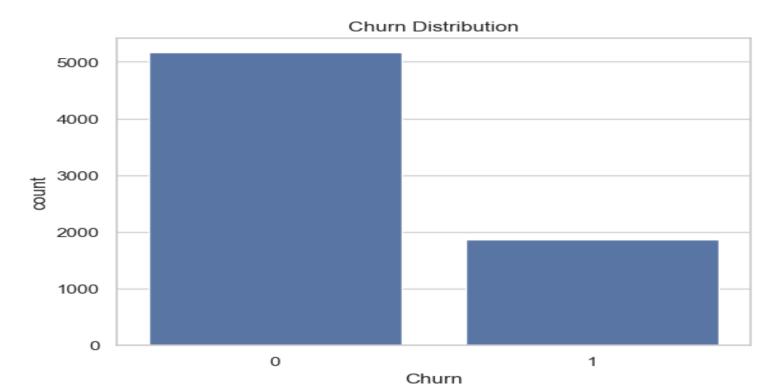
Handled missing values (TotalCharges column)

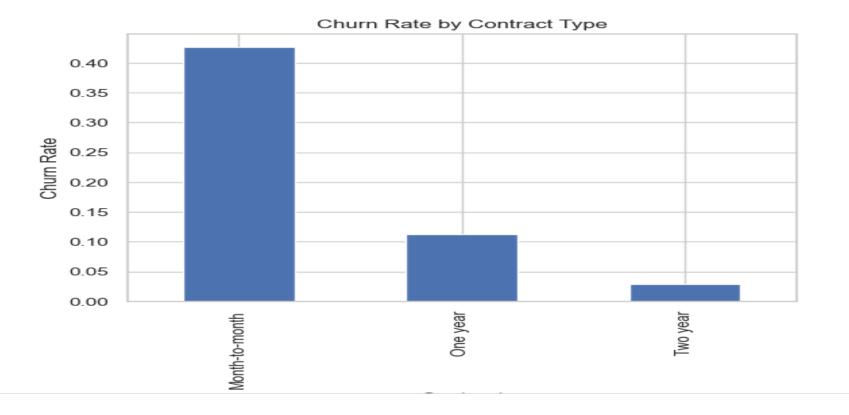
- Identified missing value in Total charges column
- Dropped the missing values
- Converted SeniorCitizen from (0/1) to Yes/No
- Encoded categorical variables for modeling
- Split into 80% training / 20% testing

# **Exploratory Data Analysis (EDA)**

**Churn Rate:**  $26\% - 27\% \rightarrow 1$  in 4 customers leave annually

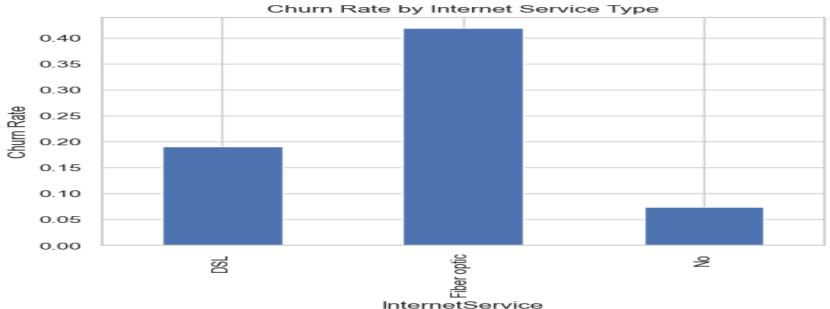
Business Impact: High churn increases costs due to expensive customer acquisition





- Month-to-month → **Highest churn** (>40%)
- Long-term contracts → Lower churn (<15%)</li>
- Recommendation: Offer incentives for long-term plans

## **Internet Service Type:**



Fiber optic users churn more than DSL users

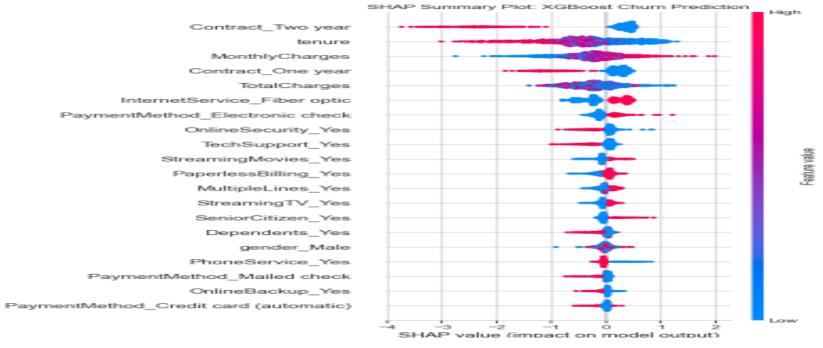
Possible reasons: Price sensitivity, competition, service issues

# Model Comparison

Мо	del Accuracy Prec	ision	Recall	F1-sco	re	
0	Random Fores	0.7853	359 0.	632353	0.459893	0.532508
1	XGBoos	0.7370	0.	503984	0.676471	0.577626
2	XGBoos	0.7370	0.	503984	0.676471	0.577626
3	XGBoos	0.7370	0.	503984	0.676471	0.577626
4	XGBoos	0.7370	0.	503984	0.676471	0.577626
5	Logistic Regression	n 0.7867	780 0.	619355	0.513369	0.561404
6	Logistic Regression	n 0.7867	780 0.	619355	0.513369	0.561404
7	XGBoos	0.7370	0.	503984	0.676471	0.577626

XGBoost chosen for highest recall → best at catching churners.

# SHAP Feature Impact (XGBoost)



- Two-year contracts & long tenure → reduce churn risk
- High monthly charges & fiber optic internet → increase churn risk
- Paying by electronic check → increases churn risk
- Support services (online security, tech support) → reduce churn risk

## Recommendations

- Promote Long-Term Contracts: Incentives for annual or two-year plans.
- Target New Customers: Focus onboarding efforts on first 6 months.
- Address High-Bill Customers: Offer loyalty discounts for high monthly charges.
- Improve Fiber Optic Retention: Service quality improvements or bundle deals.
- Encourage Support Services: Market online security and tech support packages.

# Conclusion

- Churn rate ~26.6% → a significant business challenge.
- Key churn drivers: contract type, tenure, charges, and service type.
- Predictive model (XGBoost) enables proactive churn prevention.
- Data-driven retention strategies can reduce churn and increase profitability.

