This project focuses on Customer Churn Prediction using Deep Learning (Neural Networks). It aims to identify customers who are likely to stop using a service based on various factors like tenure, charges, and internet service type.

Project Overview

- Dataset: The dataset contains customer demographics, service details, and payment methods.
- Goal: Build neural network models to predict whether a customer will churn (leave) or stay.

Key Steps

1. Data Preprocessing

- Handled Missing Values: TotalCharges was converted to numeric format, and missing values were replaced with 0.
- Feature Selection: Key features like tenure, MonthlyCharges, and TotalCharges were selected.
- Label Encoding: The target column Churn was encoded into binary values.

2. Data Visualization

- Churn Distribution: A pie chart was plotted to show the percentage of customers who churned vs. those who stayed.
- Internet Service Distribution: A count plot visualized different types of internet services used.

3. Model Building

Three different deep learning models were trained:

- 1. Model 1: A basic neural network using only tenure as input.
- 2. Model 2: Similar to Model 1 but with Dropout layers to prevent overfitting.
- 3. Model 3: A multi-feature model trained on tenure, MonthlyCharges, and TotalCharges.

Results & Evaluation

- Confusion Matrices: Used to analyze the classification performance.
- Accuracy Trends: Accuracy vs. epochs plots were generated for each model to compare training progress.

Future Improvements

- Include categorical features (e.g., Contract Type, Payment Method) using one-hot encoding.
- Experiment with more complex architectures like LSTMs or Transformer-based models.
- Use hyperparameter tuning to optimize learning rates and neuron configurations.