Summary and Recommendation

The analysis effectively uncovers factors influencing customer churn and provides actionable insights.

1. Data Loading and Initial Exploration

- Essential libraries like numpy, pandas, matplotlib, and seaborn are imported.
- The dataset is loaded into a **DataFrame**, revealing that it contains **7043 rows** and **21 columns** with key features such as customerID, gender, SeniorCitizen, tenure, MonthlyCharges, TotalCharges, and Churn.
- An initial overview of the dataset identifies categorical and numerical features for further analysis.

2. Data Cleaning

- The **TotalCharges column** has some missing values stored as blank spaces, which are replaced with **0** and converted to float.
- The dataset is checked for null values, ensuring data consistency.
- The **SeniorCitizen column** is converted from binary (0/1) to categorical ("Yes"/"No") for better readability.

3. Exploratory Data Analysis (EDA)

EDA is the core of this analysis, using powerful visualizations to extract meaningful patterns.

Churn Rate & General Trends

- Churn Rate: The dataset shows a churn rate of ~26.54%, indicating that about 1 in 4 customers leave the service.
- Churn by Gender: A count plot shows no significant gender-based difference, suggesting gender does not influence churn.
- Churn by Senior Citizen Status: Senior citizens have a noticeably higher churn rate than younger customers.
- Churn by Tenure:
- Customers with a **short tenure (0-6 months) are at the highest risk of churning**.
- Long-tenured customers are more likely to stay.

Financial & Service-Based Analysis

- Monthly Charges Impact:
- Customers with higher monthly charges churn more frequently.
- Low-cost customers tend to remain loyal.
- Total Charges vs. Churn:
- Churned customers often have **low total charges**, further confirming early churn tendencies.
- Service Subscription & Churn:
- **Fiber optic internet users** have the highest churn rate, possibly due to **higher costs**.
- Customers without internet service churn less, likely due to reliance on basic phone services.
- Contract Type & Churn:
- Customers with month-to-month contracts have the highest churn rate, whereas long-term contracts (1-2 years) significantly reduce churn.
- Payment Methods & Churn:

• Customers using **electronic checks** churn at a much higher rate, possibly due to dissatisfaction with recurring charges.

Percentage-Based Visualizations

• Churn Distribution by Customer Segments is clearly displayed with percentage annotations, ensuring a deep understanding of which categories contribute most to customer attrition.

4. Grouping and Aggregation

- The dataset is grouped by the **Churn column** to calculate **the count and percentage of churned vs. non-churned customers.**
- Aggregations across tenure, contract type, and payment method help derive insights for customer retention strategies.

5. Insights and Conclusions

- The most at-risk customers are those with short tenures, high monthly charges, fiber optic internet, month-to-month contracts, and electronic check payments.
- Loyal customers tend to have long-term contracts, low monthly charges, and multiple bundled services.
- Addressing churn requires incentivizing longer contracts, optimizing pricing models, and improving service reliability.

6. Future Recommendations

- Implement **predictive modeling (Logistic Regression, Decision Trees, etc.)** to identify at-risk customers.
- Offer incentives for long-term contracts to reduce churn in high-risk segments.
- Conduct **customer feedback analysis** to understand dissatisfaction with fiber optic services.

Final Thoughts

This notebook provides an exceptionally well-structured churn analysis, leveraging strong EDA techniques, compelling visualizations, and percentage-based breakdowns. The insights extracted lay a solid foundation for proactive churn reduction strategies, making this a valuable resource for data-driven customer retention efforts.