

# Telecom Customer Churn Analysis

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IST-718 Big Data Analytics

# Agenda

## Customer Churn in Telecom

- Problem Background
- Data

## Exploratory Analysis

- Demographic Data
- Customer Relationship Data

## Models

- Logistic Regression
- Random Forest
- Support Vector Machine

# Customer Churn in Telecom

- Subscription Based Business:
  - Monthly,
  - 1-year,
  - 2-year contracts
- Revenue and long-term growth
- Data
  - Telecom Customer Churn Prediction dataset from Kaggle
  - Telecom Company in California
  - Customer Data Q2 2022
  - Population by zip code
  - <https://www.kaggle.com/datasets/shilongzhuang/telecom-customer-churn-by-maven-analytics/data>



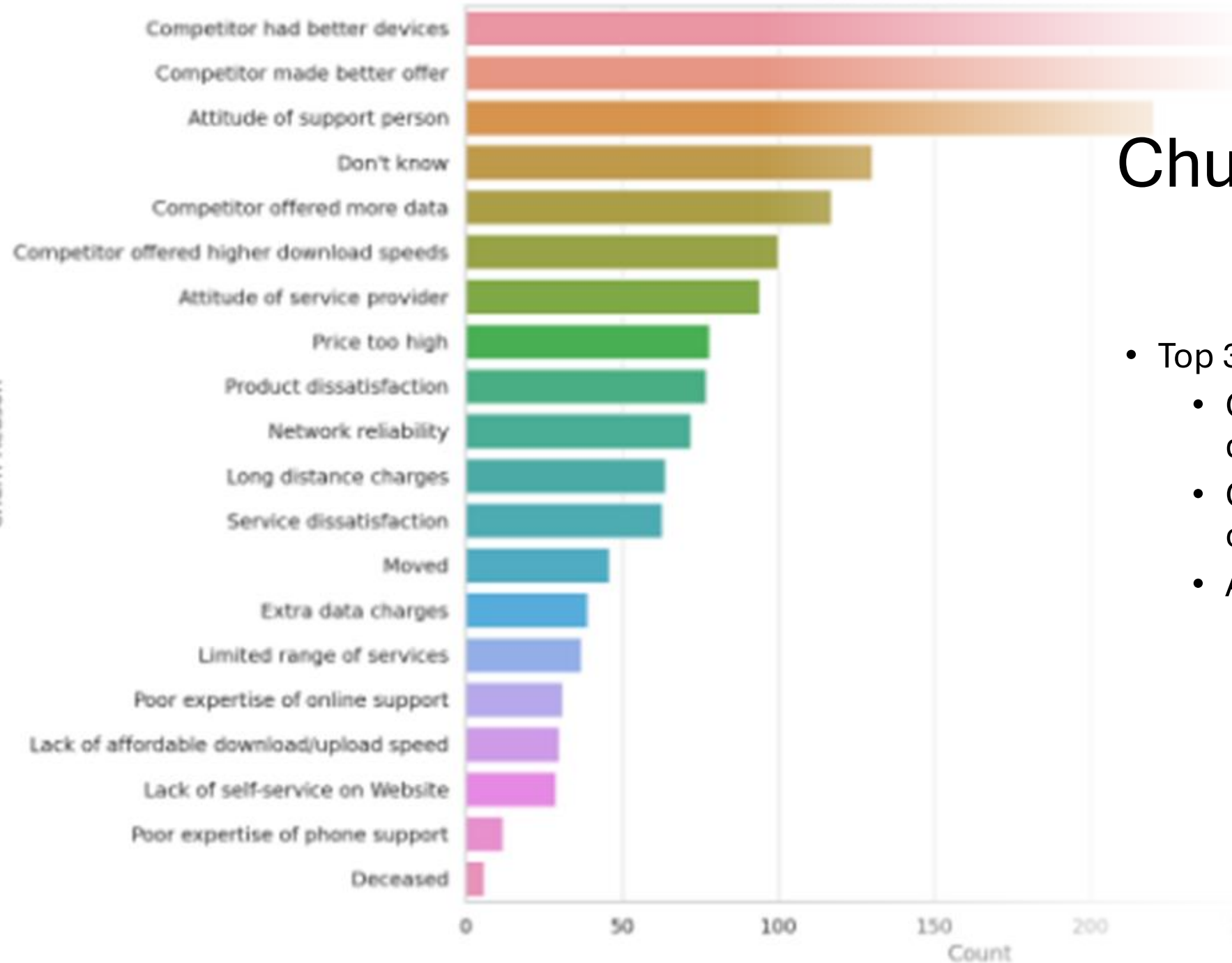
# Exploratory Data Analysis

Distributions

Correlations with Churn

Distribution of Churn Reasons

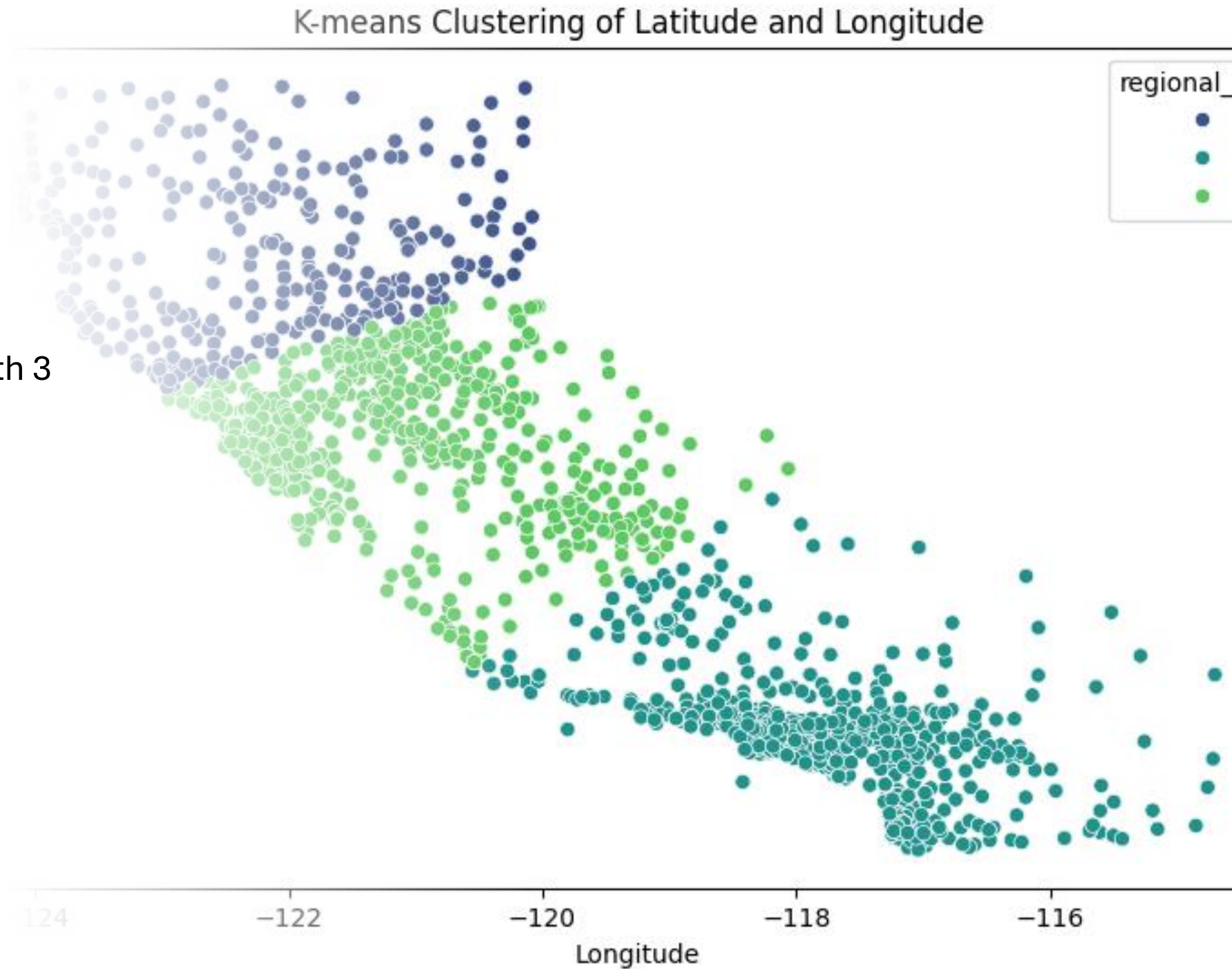
# Churn Reason



- Top 3
  - Competitor has better devices
  - Competitor made better offer
  - Attitude of support person

# Location

- K-Means Clustering with 3 clusters
  - Northern CA
  - Central CA
  - Southern CA







# Predictive Models

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Logistic Regression

Random Forest Classifier

Support Vector Machine

# Models Test Accuracy

## Logistic Regression

- Unbalanced Data: 83.7%
- Down-sampled Data: 78.1%
- Up-sampled Data: 78.9%

## Random Forest

- Unbalanced Data: 84.6%
- Down-sampled Data: 79.2%
- Up-sampled Data: 81.5%

## Support Vector Machine

- Unbalanced Data: 84.6%
- Down-sampled Data: 77.1%
- Up-sampled Data: 76.3%



## Best Model

- Support Vector Machine: 84.6% Test Accuracy
- Unbalanced Data

Confusion Matrix

True labels	Churned	Stayed
Churned	241	132
Stayed	85	951
Predicted labels		

# Final Thoughts

- Best model was achieved with Support Vector Machine
  - 241 of the 373 churned customers in the test set were predicted as churned
- Recommendations:
  - Pricing Strategy Reevaluation
  - Enhancing Customer Engagement
  - Geographical and Demographic Targeting
  - Service Improvement

