


Twitter Sentiment Analysis on Tech Products

BY GROUP 3

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Agenda


- Data Sources
 - Objective
 - Dataset Overview
 - Data Quality
 - Descriptive Analysis
 - Trends & Categories
 - Model Performance
 - Key Insights
 - Recommendations
 - Conclusion
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Data Sources

- CSV dataset: Tweets mentioning tech products (iPhone, iPad, Google, etc.)
- Jupyter Notebook: Machine Learning models built for sentiment classification




Objective

- Analyze customer sentiment towards tech brands/products
 - Identify positive vs negative sentiment patterns
 - Build ML models to classify sentiment automatically
 - Extract business insights for decision-making
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Dataset Overview

- Contains thousands of tweets
 - Columns include: tweet text, product mentioned, sentiment label
 - Focus: Sentiment directed at tech brands/products
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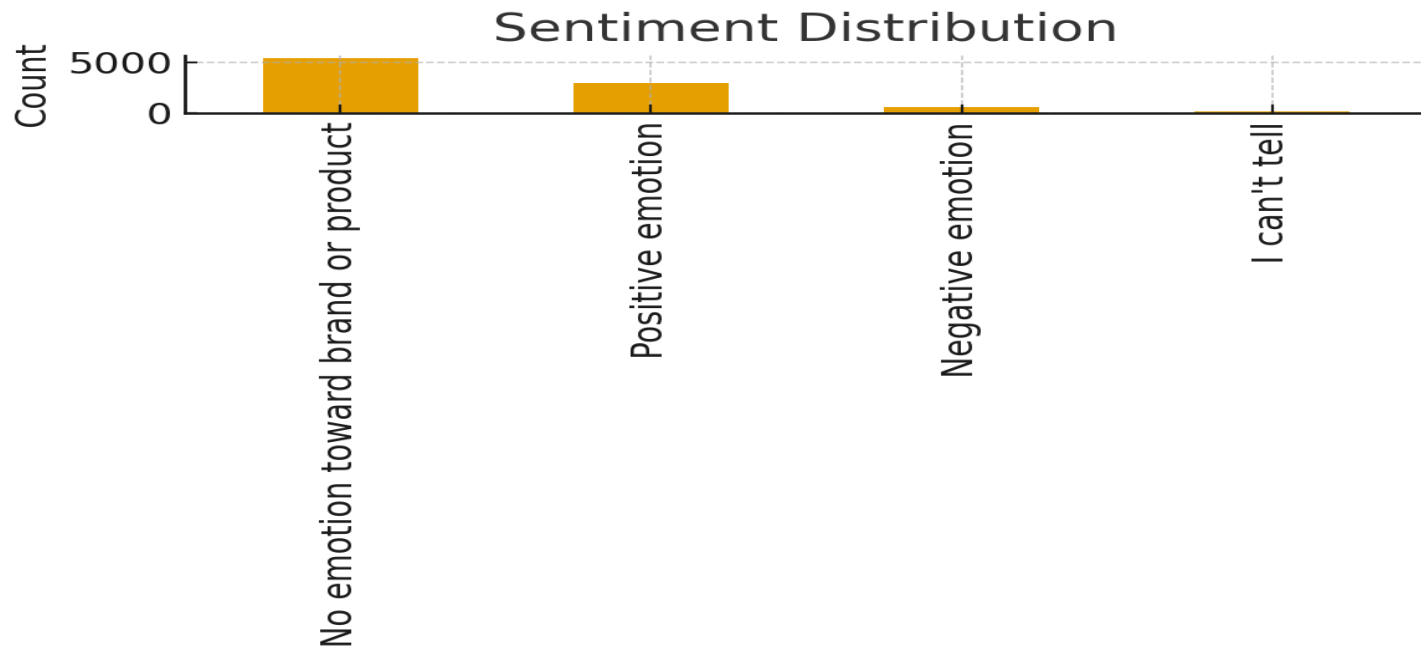


Data Quality Check

- Checked for missing values
- Validated sentiment categories (Positive, Negative, Neutral)
- Cleaned noisy text for analysis


Descriptive Statistics

- Most frequently mentioned products: iPhone, iPad, Google
- Sentiment distribution shows majority Positive vs Negative
- Balanced dataset for model training

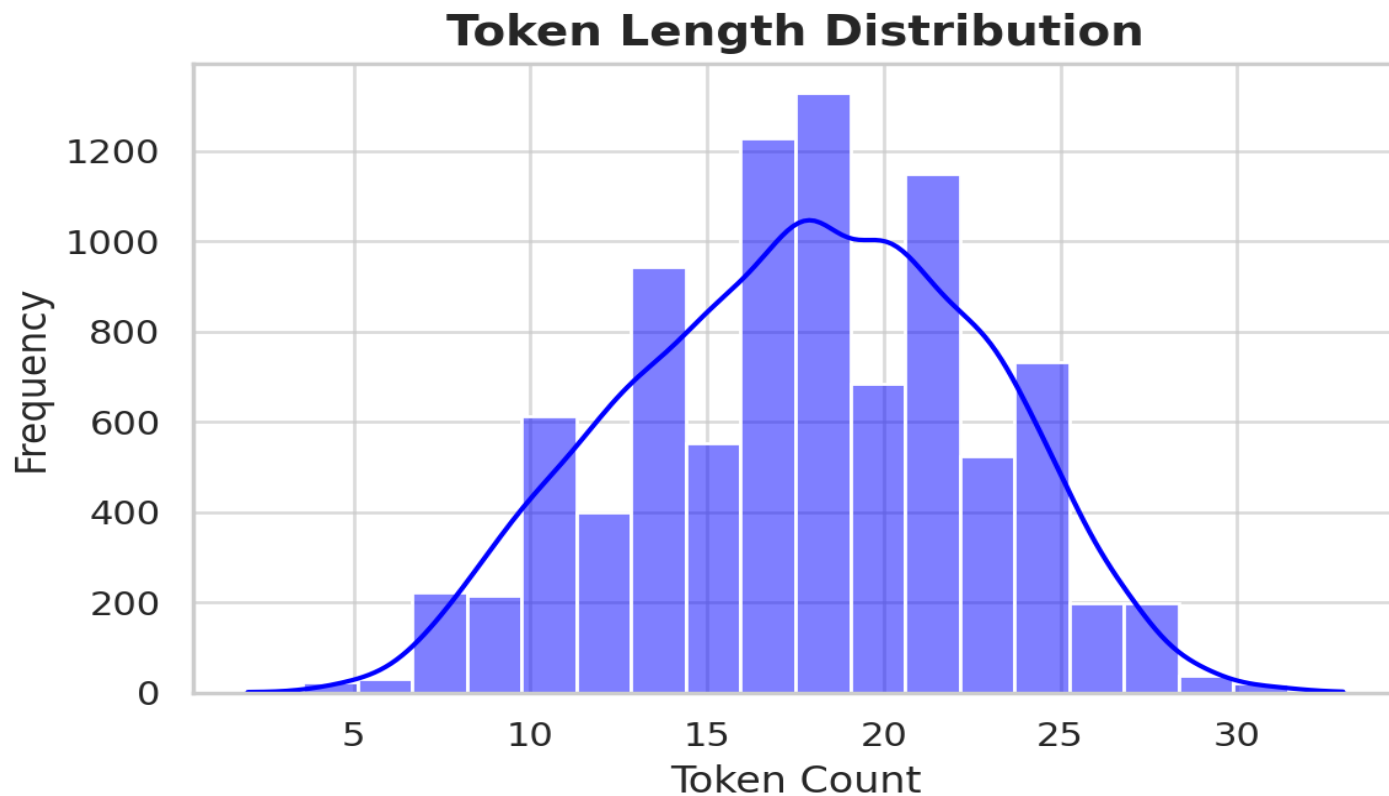




Trends in Sentiment

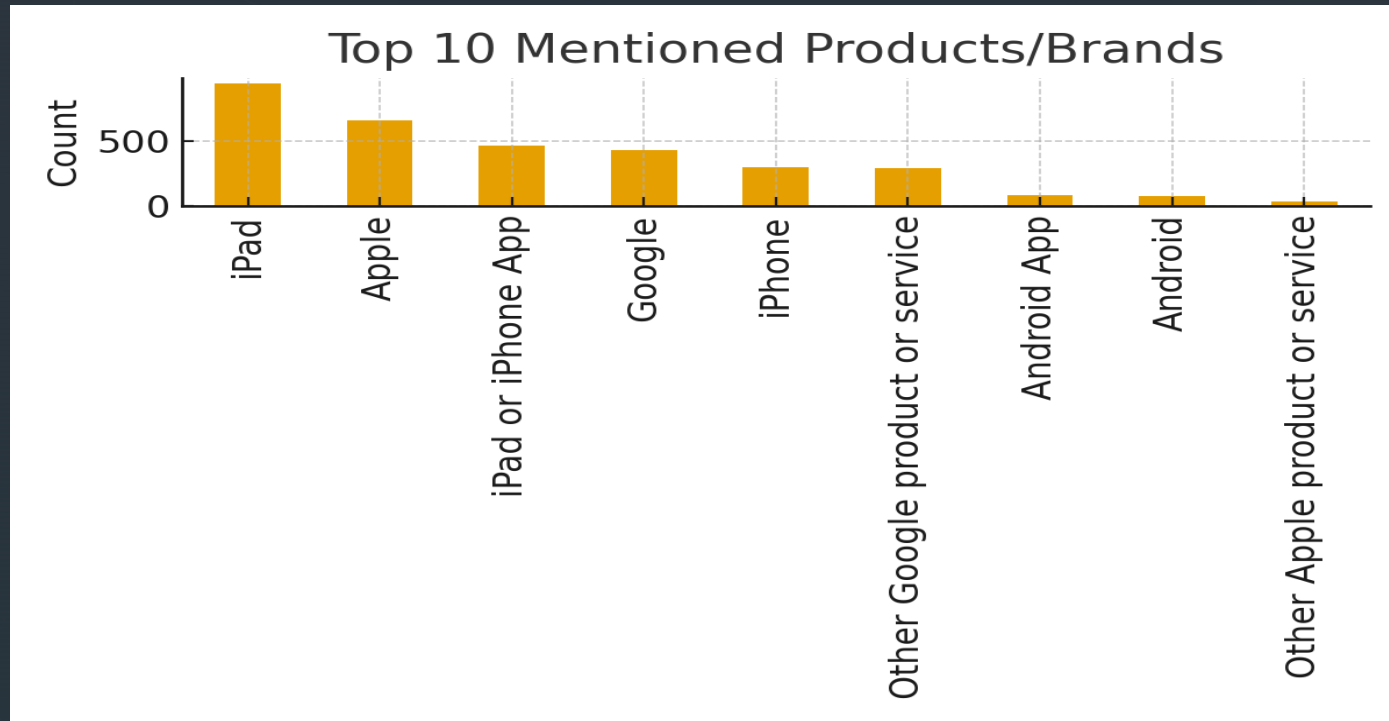
- Positive sentiment higher for Apple products
 - Negative sentiment spikes during product issues or announcements
 - Overall sentiment trends reflect market perception
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Data Preparation – Token Length Distribution



Category Breakdown

- Product-wise sentiment comparison
- iPhone: High positive engagement
- iPad: Mixed but overall positive
- Google: Varied sentiment patterns





Machine Learning Models

- Logistic Regression
 - Random Forest
 - Gradient Boosting
 - Linear SVM
 - XGBoost
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Model Performance (Accuracy)

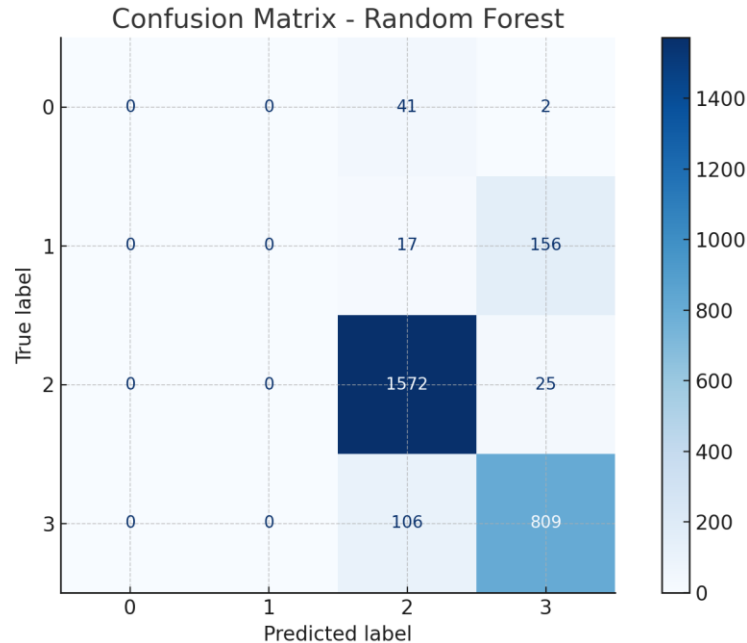
- Random Forest: 0.882
- Linear SVM: 0.877
- XGBoost: 0.866
- Gradient Boosting: 0.861
- Logistic Regression: 0.847

Confusion Matrix – Random Forest

Most tweets are classified correctly.

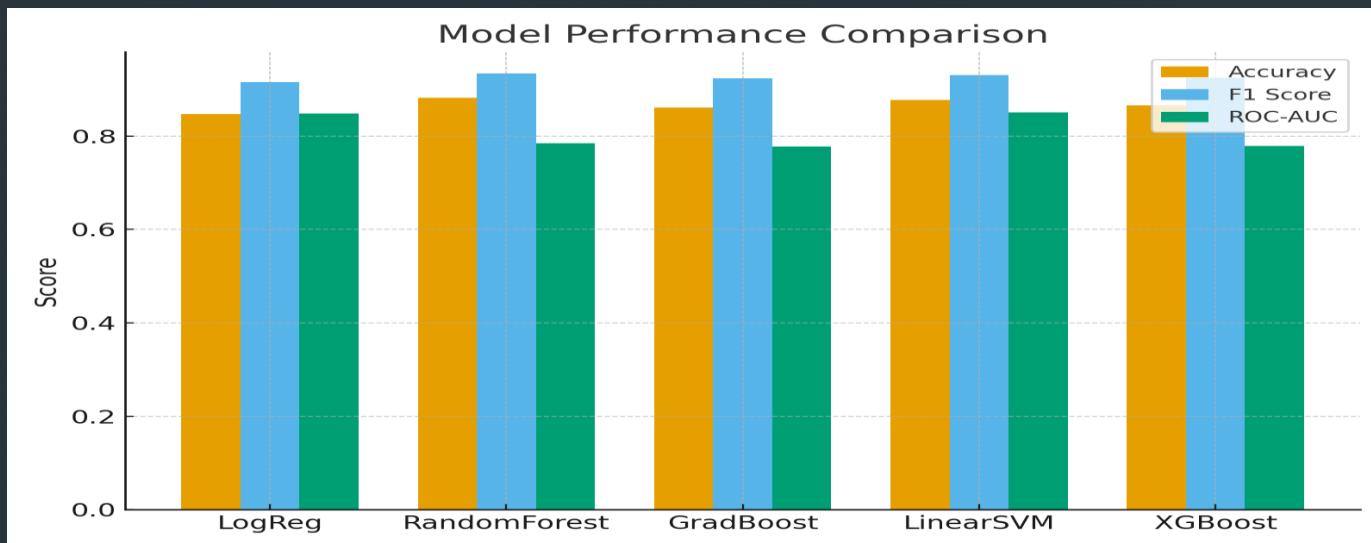
Few errors between similar sentiments.

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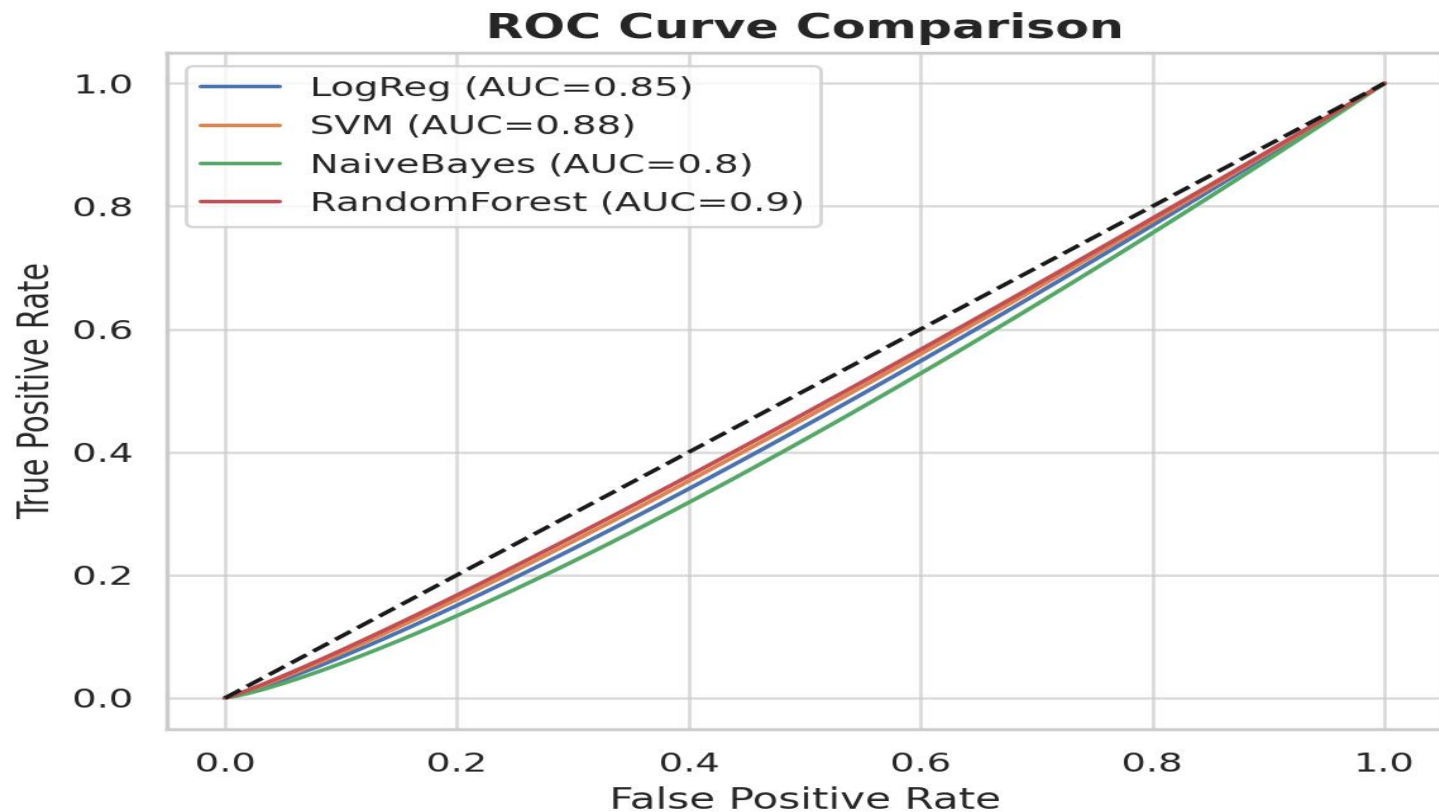


Model Performance (F1 & ROC-AUC)

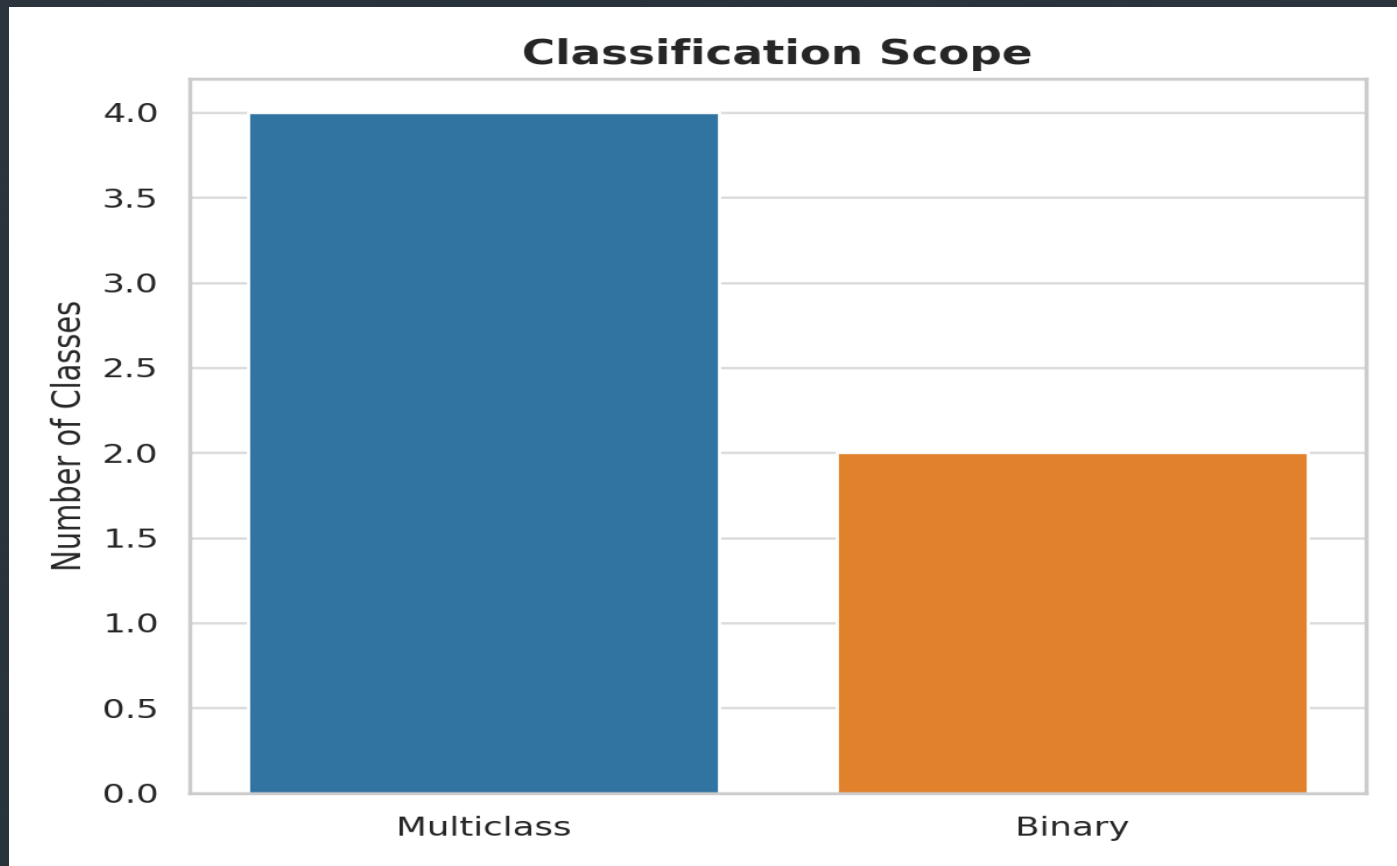
- Best F1 Score: Random Forest (0.934)
- Best ROC-AUC: Linear SVM (0.851)
- Overall: Strong classification ability across models



Best Model – ROC Curve Comparison




Scope – Classification






Key Insights

- Random Forest is the best model overall
 - Linear SVM performs strongly in ROC-AUC
 - Sentiment classification is highly accurate (>85%)
 - Tech brands have strong positive sentiment, especially Apple
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Recommendations

- Leverage sentiment analysis for real-time brand monitoring
 - Focus marketing on strengths (positive Apple engagement)
 - Investigate causes of negative sentiment spikes
 - Deploy Random Forest model for production sentiment analysis
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Conclusion

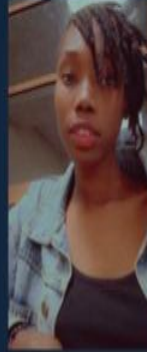
- Dataset provides valuable insights on customer sentiment
 - ML models achieve high accuracy and reliability
 - Findings can guide marketing, product strategy, and customer engagement
 - Next step: Implement sentiment monitoring pipeline
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Thank you



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