

# Twitter Sentiment Analysis for Apple Products

By group 11 DSFT-13



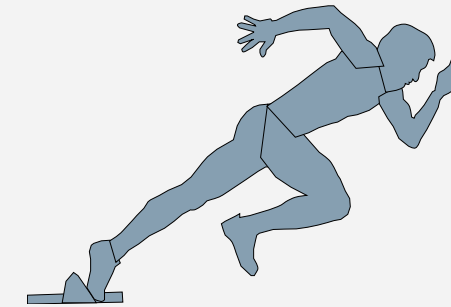
# Transforming How We Listen to Customers



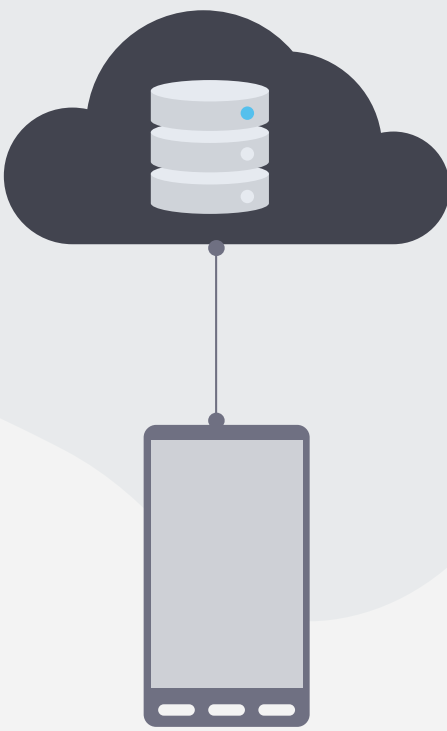
- Automated system to classify Twitter sentiment in real-time



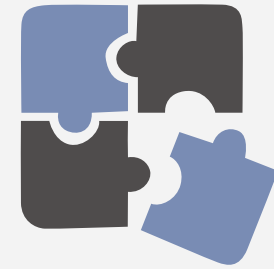
- Focus on catching customer complaints quickly



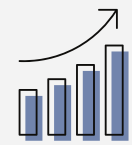
- Enables faster response to product issues
- Deploy the Classifier



# Why We Need This Solution



## The Challenge We Face



- Thousands of daily mentions about Apple products on Twitter

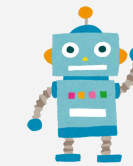


- Manual sentiment analysis is slow and inconsistent



- Critical complaints get missed or delayed

## Our Automated Solution



- Real-time Twitter sentiment classification



- Enables faster response to issues



- Focus on catching complaints automatically

- **Business Value:** Improved customer satisfaction & product quality





# What We Set Out to Achieve





- **Primary Objective:** Catch customer complaints automatically

## OUR TARGETS:

-  **>45% Negative Recall | Achieved: 50%**
  - Catch nearly half of all complaints

-  **Handle Imbalanced data**
  - Focus on rare but critical feedback

-  **Real-Time Preprocessing**
  - Instant analysis of thousands of tweets

-  **Deliver Actionable Insights**
  - Product teams can trust and act on results



# Our 5-Step Analysis Journey



## Feature engineering

## Handling Imbalance

1

- Transforming text into signals
- Identifying key words
- Creating sentiment indicators

3

- Special focus on rare but critical complaints
- Addressing the 7% negative tweet challenge

5

## Data Preparation

- Cleaning 9,093 tweets
- Removing special characters
- Standardizing text format.

2

## Model Selection

- Choosing the right approach for sentiment
- Testing multiple algorithms
- Selecting best performer for our data

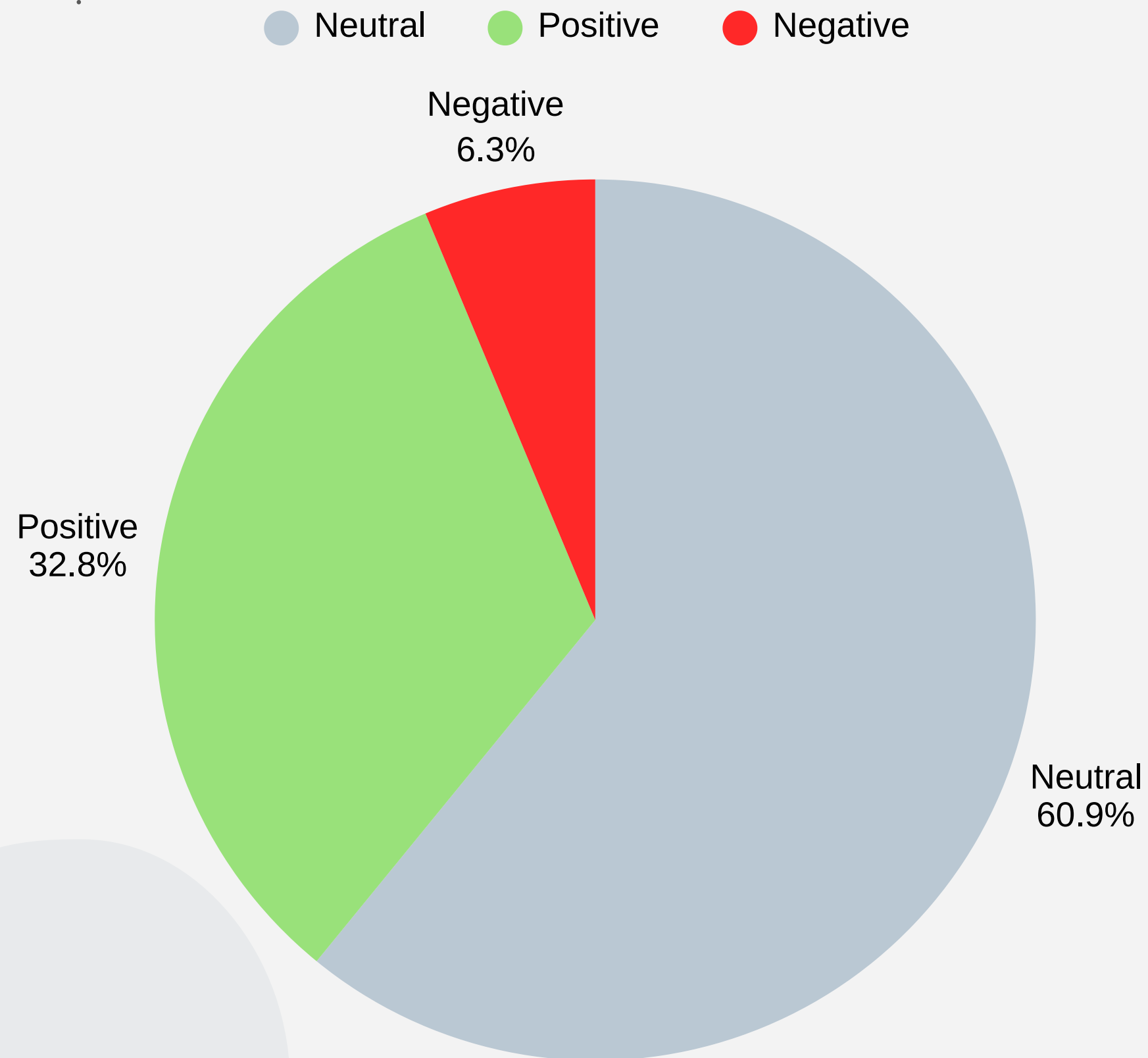
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## Performance Evaluation

- Measuring against business targets
- Evaluating recall and accuracy
- Validating real-world performance

# The critical 6%: where the Business value lives

## Sentiment Proportion



## Key Insights:

- 61% Neutral tweets
- 33% Positive feedback
- 6% Negative = Customer Complaints

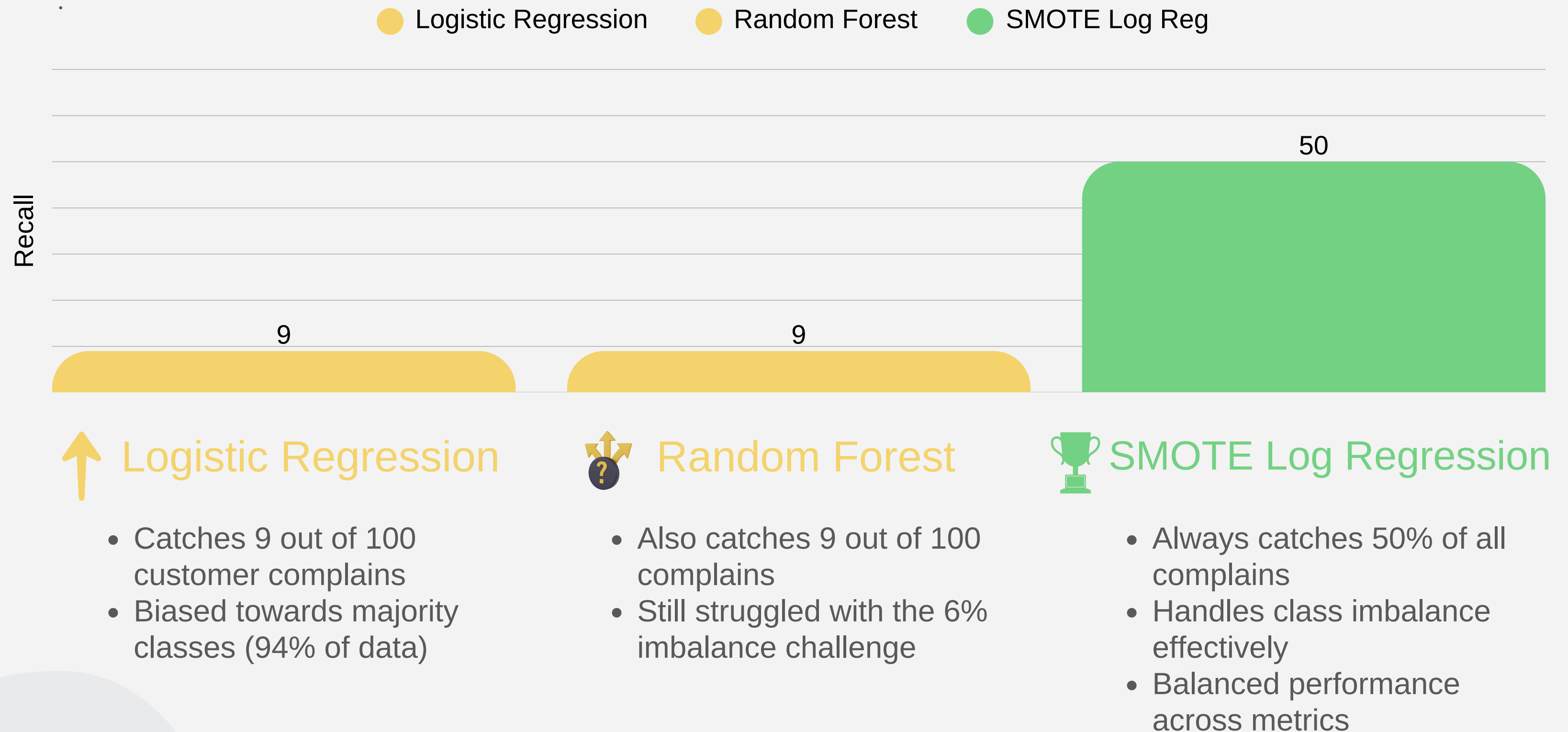
## Business Impact:

- The rare 570 negative tweets (6% of total) represent our biggest opportunity for customer satisfaction improvement

## Data:

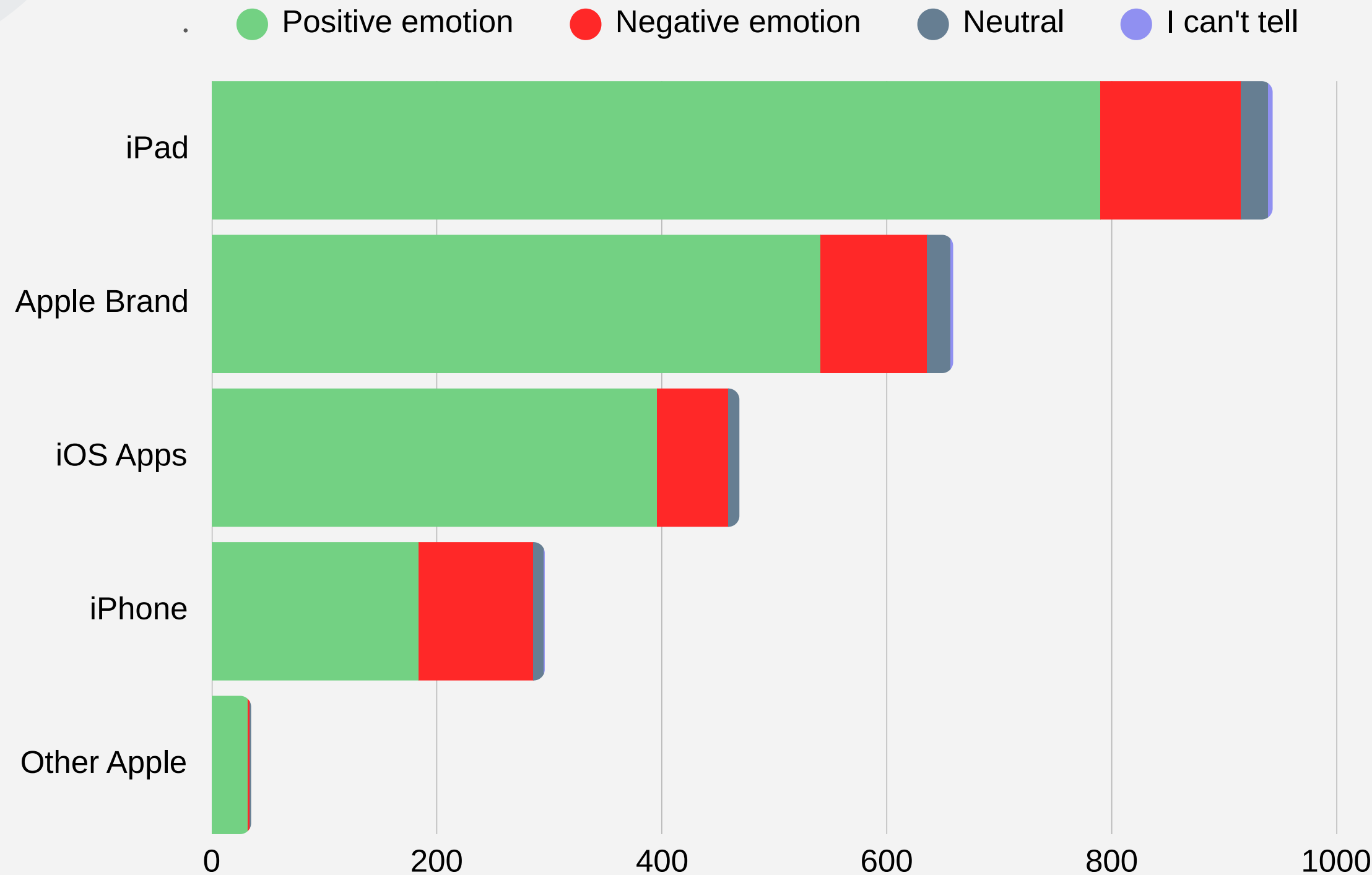
- 9,093 human-labeled tweets from CrowdFlower

# Choosing the Right Approach: Why SMOTE Won



# Business Impact: Targeted Product Insights

## Products vs. sentiments



## Priority actions:



### iPhone:

- Urgent attention needed
- 34% of tweets are complaints
- Focus on core functionality



### iPad:

- Monitor and maintain
- High volume, moderate complaints
- Continue current quality



### iOS Apps:

- Success story
- 84% positive sentiment
- Expand successful features

- *Targeted product insights drive high-impact improvements in customer satisfaction.*



# Recommendations: From Insights to Impact



## Immediate Actions

Integrate real-time monitoring  
Set up automated alerts  
Team training on dashboard



## Strategic Focus Areas

**iPhone:** Address core functionality issues  
**iPad:** Maintain quality, monitor trends  
**iOS Apps:** Expand successful features



## Longterm Value

Quarterly model updates  
Platform expansion (Instagram,  
Reddit)  
Product launch sentiment tracking

# Our Evolving Customer Intelligence



## Current Capabilities

Twitter sentiment  
Real-time analysis  
Product-specific insights  
50% complaint recall



## Next Phase Development

**Q1 2026:** Instagram + Reddit integration  
**Q2 2026:** Predictive analytics prototype  
**Q3 2026:** Global sentiment expansion



## Future Vision:

Multi-platform monitoring  
Predictive issue detection  
Automated response system  
Global market intelligence

# From Notebook to production: Live Demo



## Deployed & Ready to Use

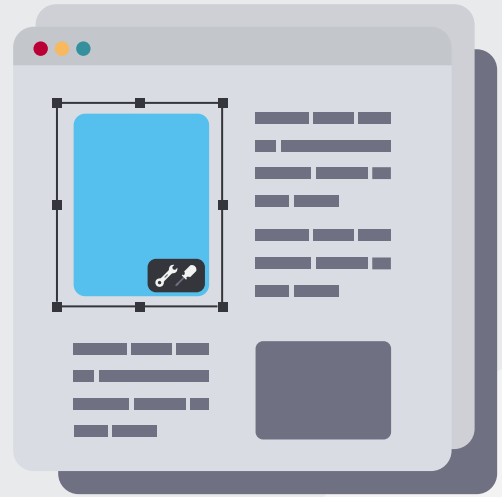
Real-time web application  
Accessible on any device  
Instant sentiment predictions

 *Link*

## Features

Test with any Apple product tweet  
See confidence scores for each prediction  
Understand model decision-making





# Thanks!

Do you have any questions?

[Github](#)

[Sentiment Classifier App](#)

