
SENTIMENT ANALYSIS

Comprehensive Report

Generated: 2026-02-18 01:34:22

Dataset: 1,000 records analyzed

Models: ALL

Output Directory: sentiment_results/

Sentiment Analysis Automation — Commercial Edition

Executive Summary

This comprehensive sentiment analysis report presents findings from **1,000 records** processed using state-of-the-art natural language processing models. The analysis reveals a **positive** overall sentiment with **82.9%** of records classified as positive. The net sentiment score stands at **+69.0%**, indicating strong positive sentiment across the dataset.

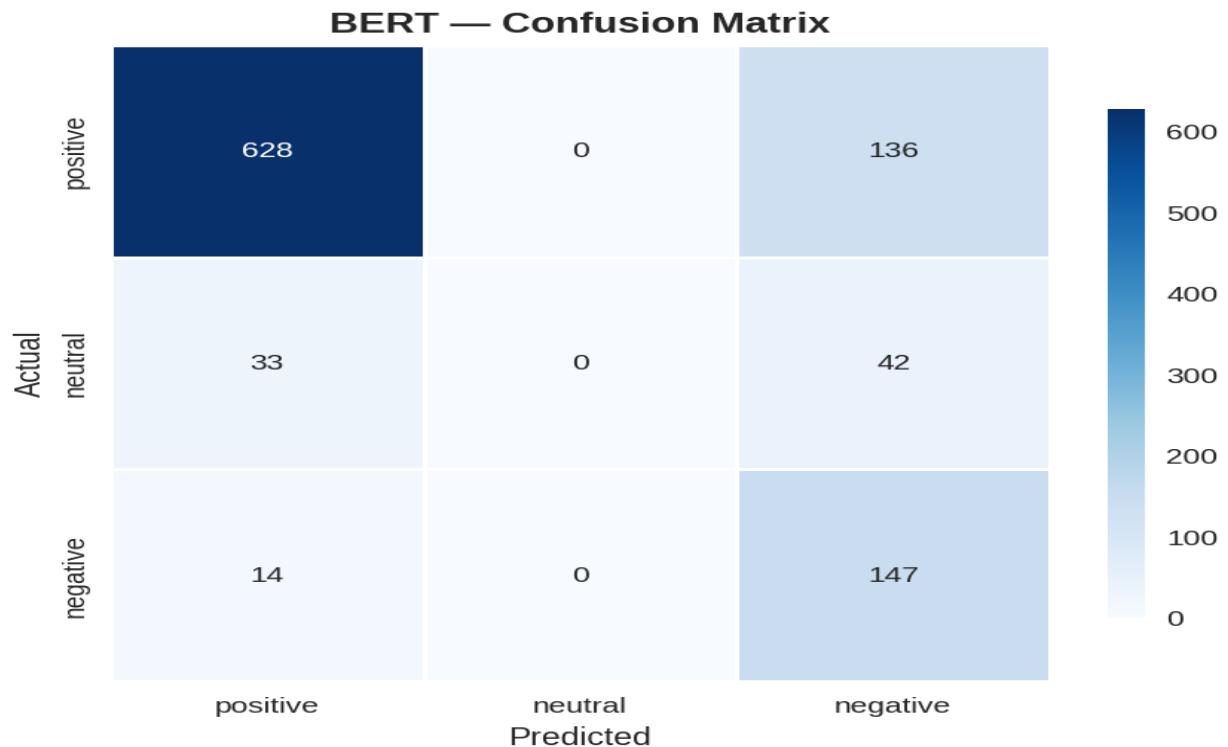
Sentiment	Count	Percentage
Positive	829	82.9%
Neutral	32	3.2%
Negative	139	13.9%

Key Insights

- **Insight 1:** The dominant sentiment is 'positive' with 82.9% of records.
- **Insight 2:** Strong positive net sentiment (+69.0%). The audience is highly satisfied.

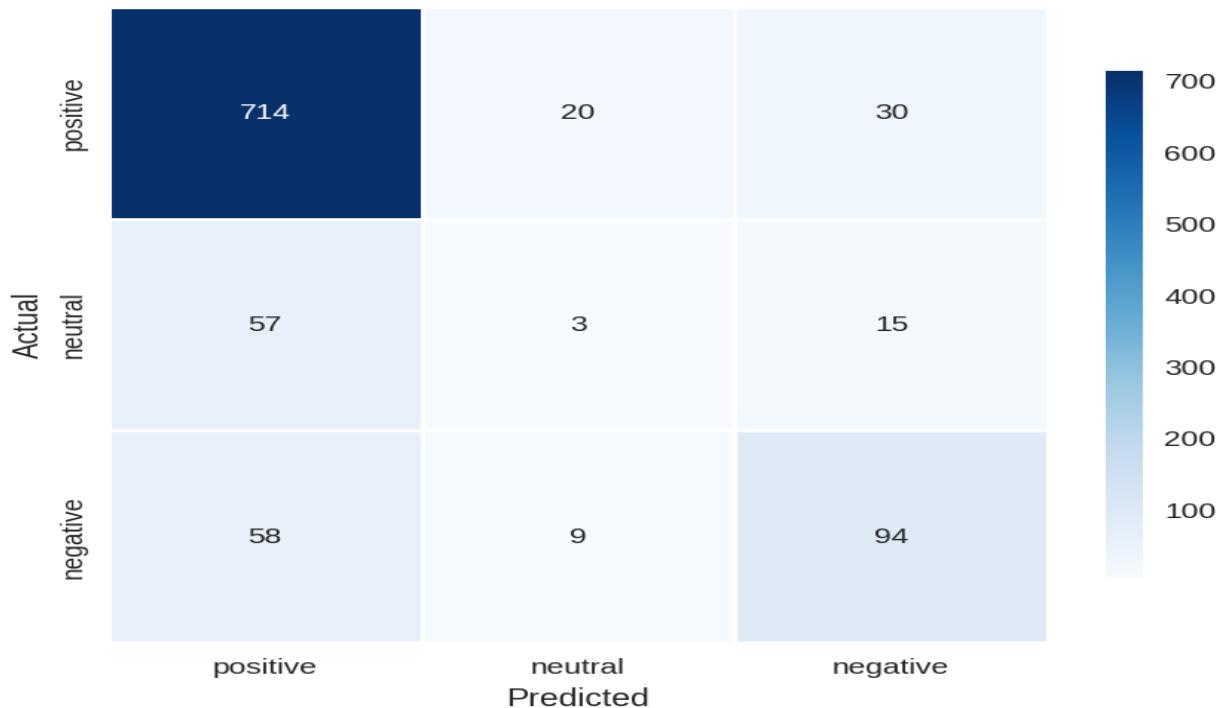
Visualizations

Confusion Matrix Bert

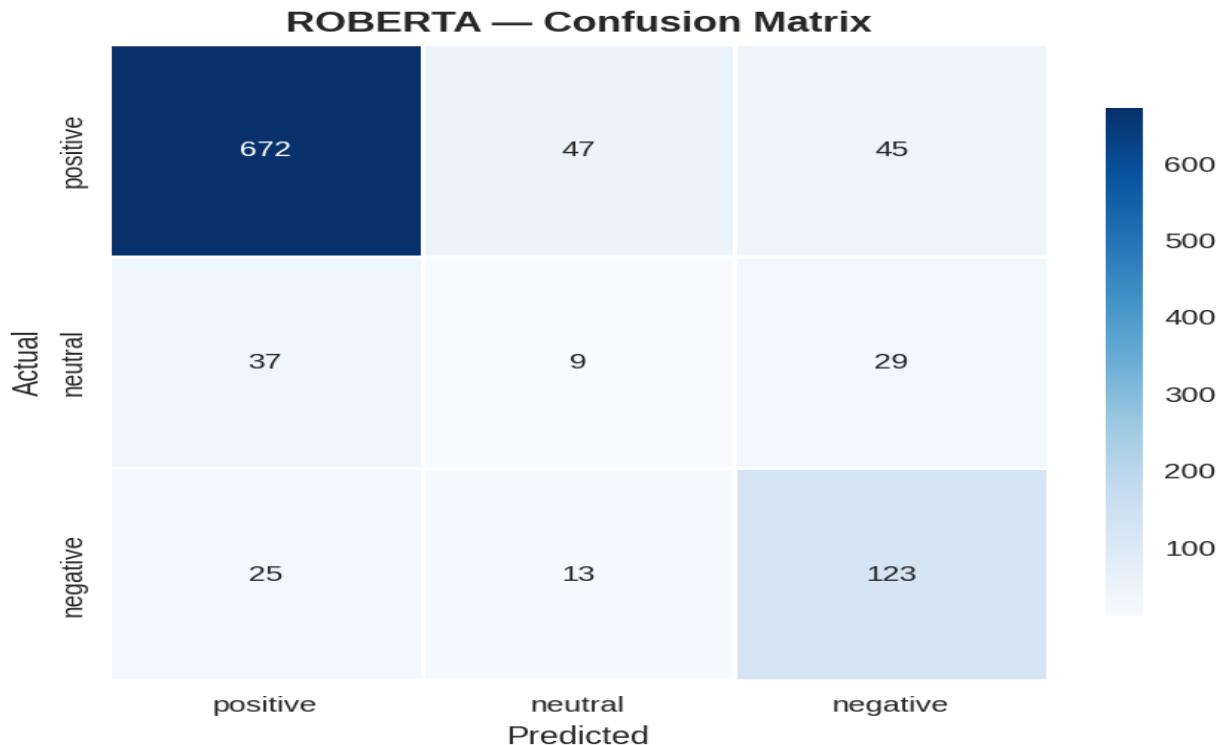


Confusion Matrix Ensemble

ENSEMBLE — Confusion Matrix

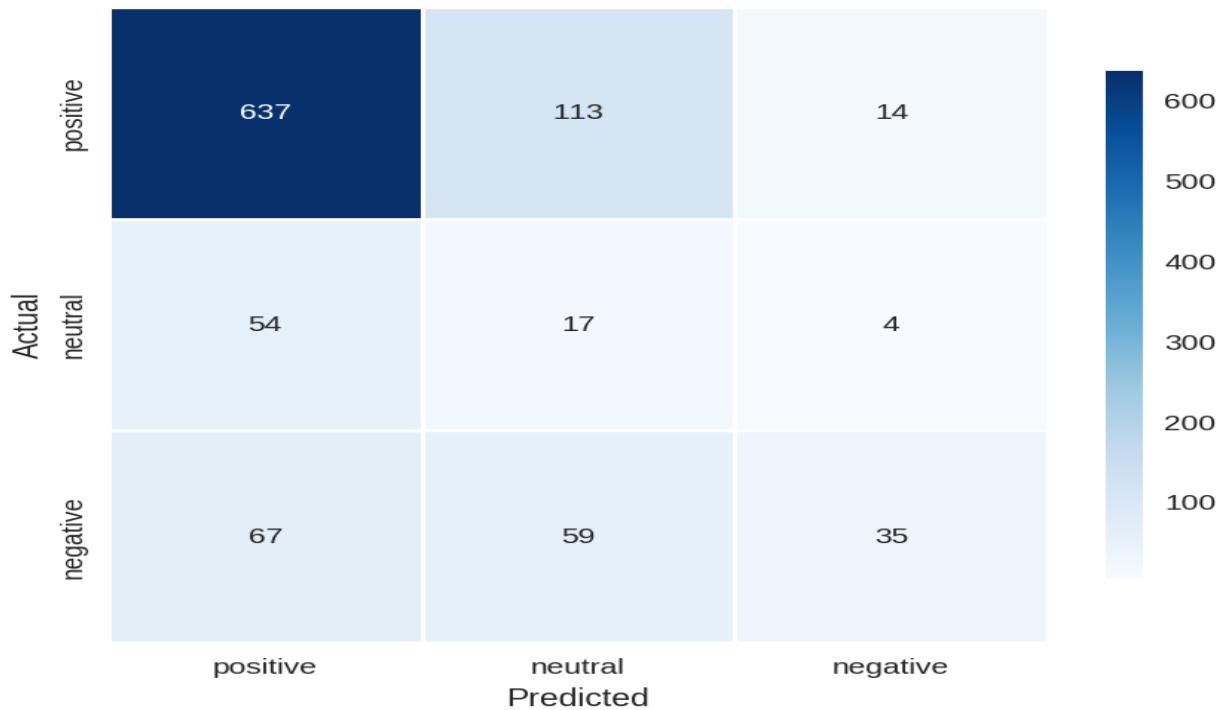


Confusion Matrix Roberta

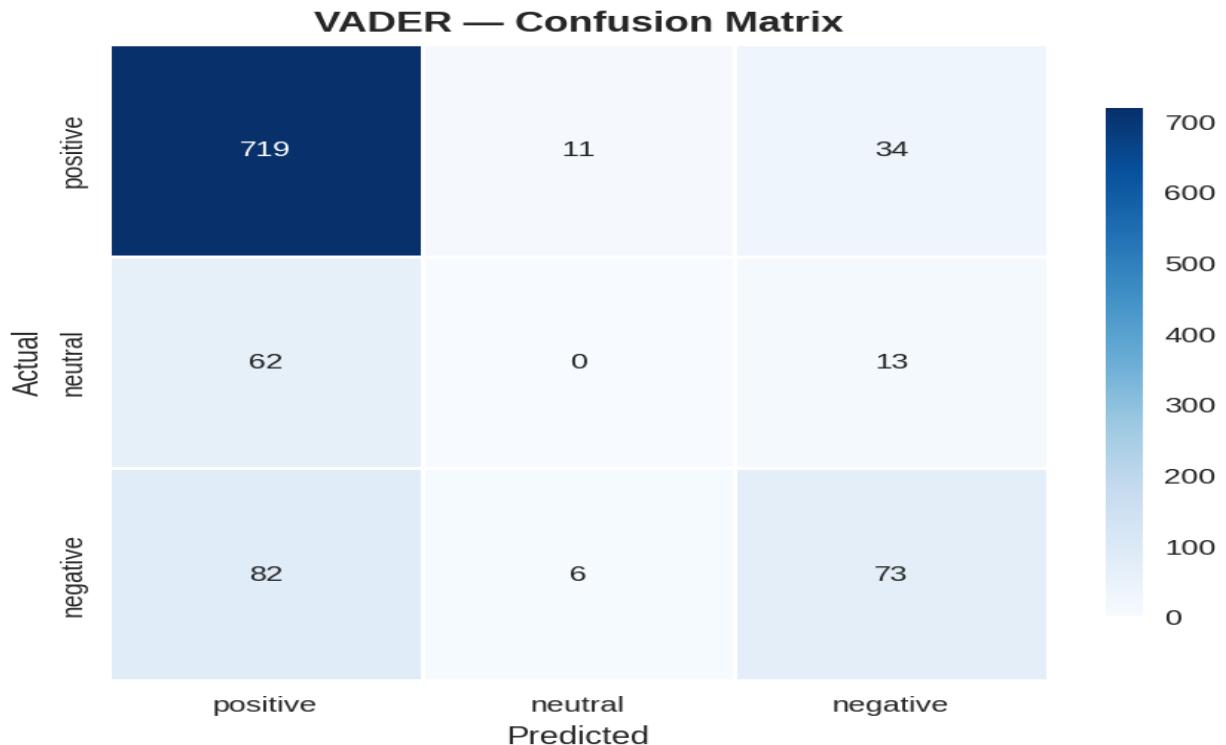


Confusion Matrix Textblob

TEXTBLOB — Confusion Matrix



Confusion Matrix Vader

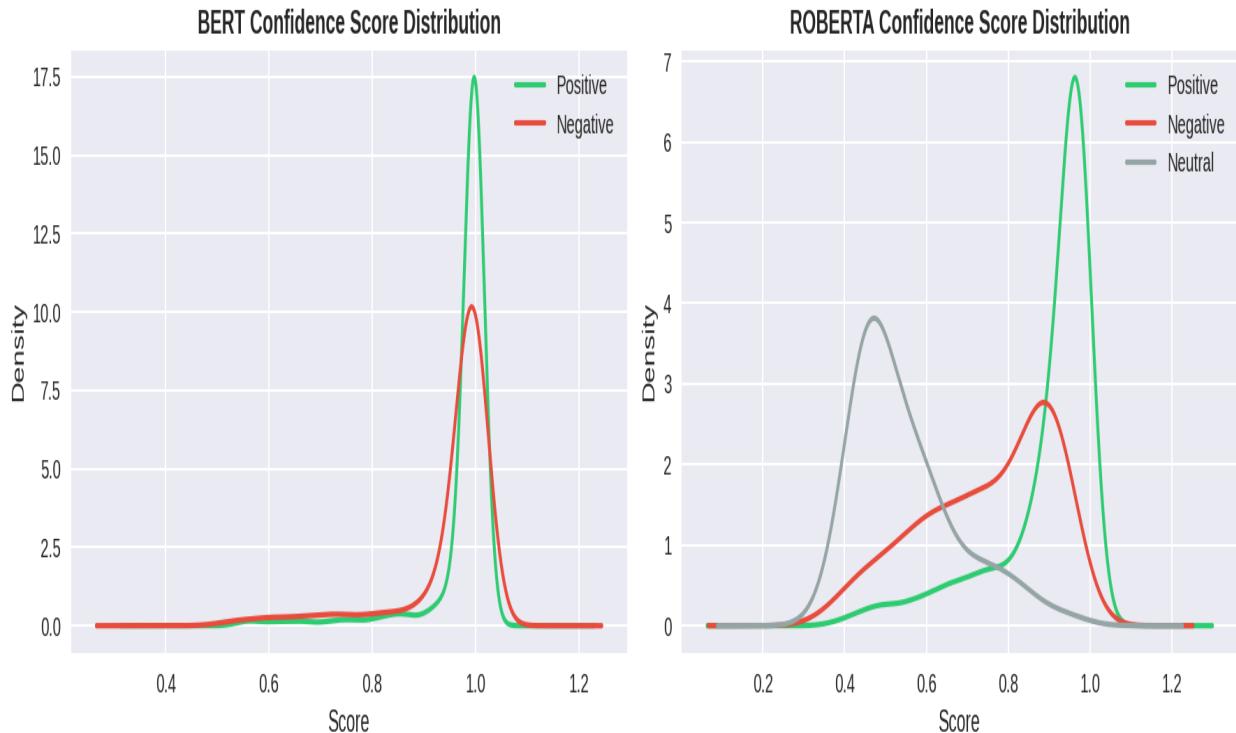


Model Comparison

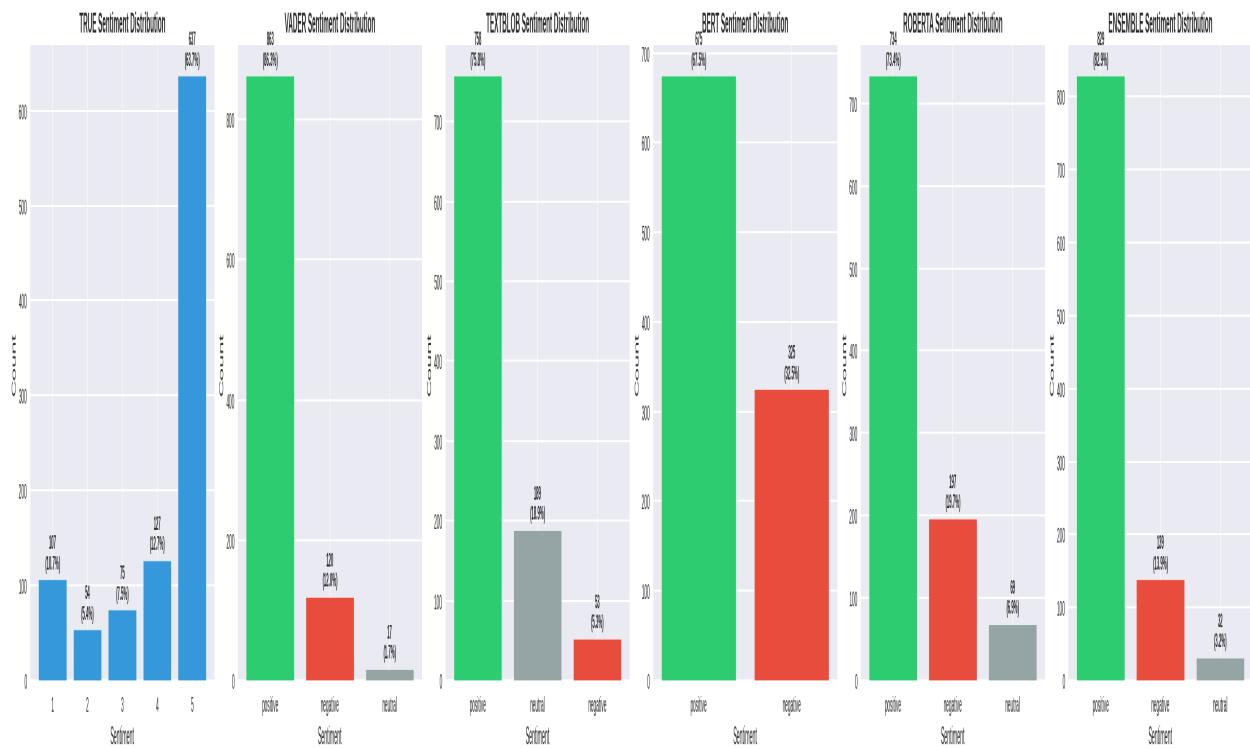
Model Performance Comparison



Score Distributions

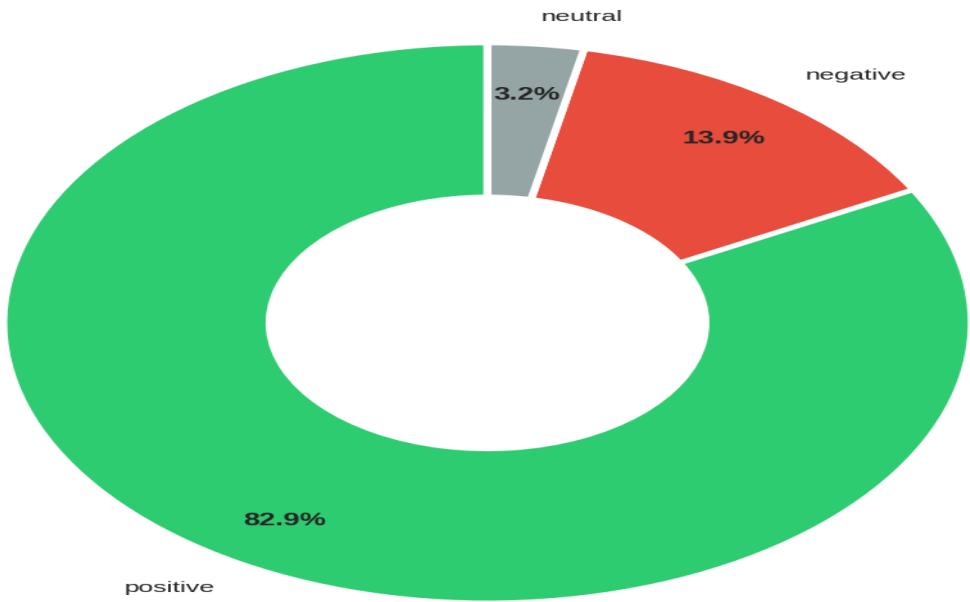


Sentiment Distribution



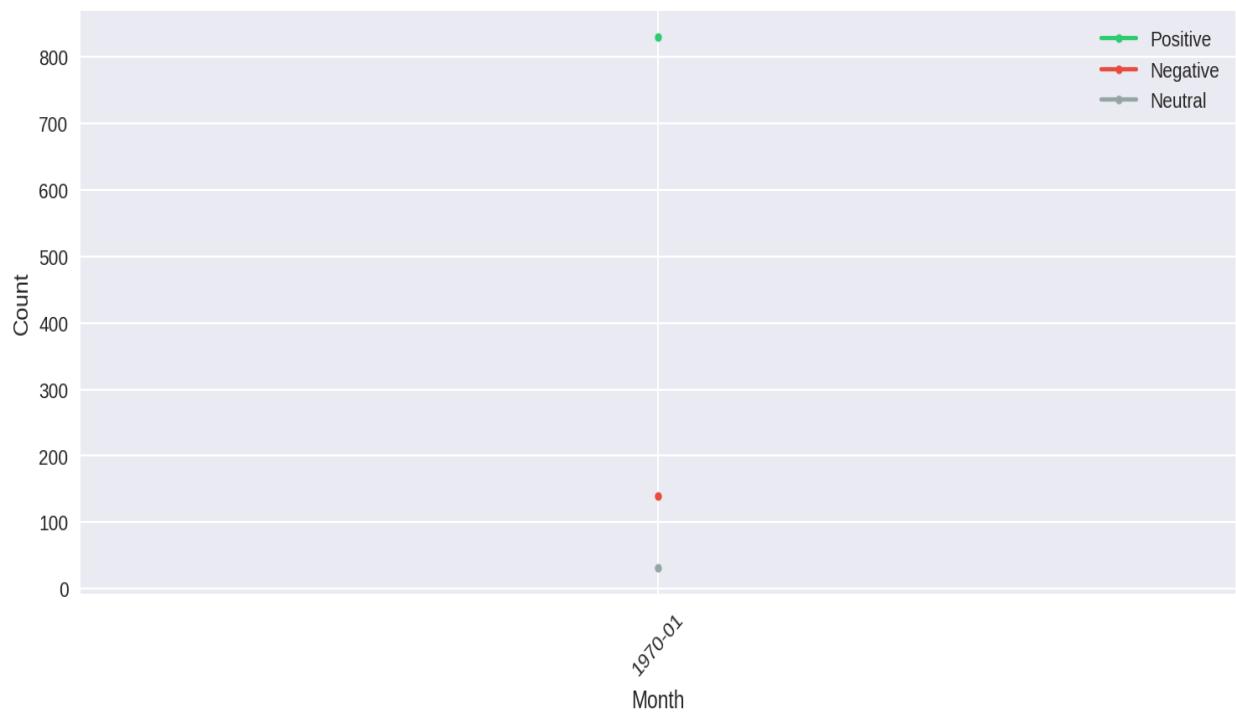
Sentiment Pie

Sentiment Distribution — ENSEMBLE



Sentiment Timeline

Sentiment Trends Over Time



Word Clouds by Sentiment

Word Clouds by Sentiment



Model Performance

Model	Accuracy	F1 Score
VADER	79.20%	0.7589
TEXTBLOB	68.90%	0.7018
BERT	77.50%	0.7642
ROBERTA	80.40%	0.8055
ENSEMBLE	81.10%	0.7900

VADER — Classification Report

```
precision recall f1-score support
```

```
negative 0.61 0.45 0.52 161
neutral 0.00 0.00 0.00 75
positive 0.83 0.94 0.88 764
```

```
accuracy 0.79 1000
macro avg 0.48 0.46 0.47 1000
weighted avg 0.73 0.79 0.76 1000
```

TEXTBLOB — Classification Report

```
precision recall f1-score support
```

```
negative 0.66 0.22 0.33 161
neutral 0.09 0.23 0.13 75
positive 0.84 0.83 0.84 764
```

```
accuracy 0.69 1000
macro avg 0.53 0.43 0.43 1000
weighted avg 0.76 0.69 0.70 1000
```

BERT — Classification Report

```
precision recall f1-score support
```

```
negative 0.45 0.91 0.60 161
neutral 0.00 0.00 0.00 75
positive 0.93 0.82 0.87 764
```

```
accuracy 0.78 1000
macro avg 0.46 0.58 0.49 1000
weighted avg 0.78 0.78 0.76 1000
```

ROBERTA — Classification Report

```
precision recall f1-score support
```

```
negative 0.62 0.76 0.69 161
```

```
neutral 0.13 0.12 0.12 75
```

```
positive 0.92 0.88 0.90 764
```

```
accuracy 0.80 1000
```

```
macro avg 0.56 0.59 0.57 1000
```

```
weighted avg 0.81 0.80 0.81 1000
```

ENSEMBLE — Classification Report

```
precision recall f1-score support
```

```
negative 0.68 0.58 0.63 161
```

```
neutral 0.09 0.04 0.06 75
```

```
positive 0.86 0.93 0.90 764
```

```
accuracy 0.81 1000
```

```
macro avg 0.54 0.52 0.53 1000
```

```
weighted avg 0.77 0.81 0.79 1000
```

Recommendations

- 1. Leverage Positive Sentiment:** The overwhelmingly positive sentiment presents an opportunity to amplify success stories through marketing campaigns and customer testimonials.
- 2. Continuous Monitoring:** Deploy this sentiment analysis pipeline as a recurring process to track trends and measure the impact of improvement initiatives.

Methodology

Data Processing: The analysis pipeline processed 1,000 text records with comprehensive preprocessing including tokenization, lemmatization, and stop-word removal. Text length was standardized to 512 characters for optimal model performance.

Models Deployed: Multiple state-of-the-art NLP models were employed:

- **VADER** (Valence Aware Dictionary and sEntiment Reasoner) — Rule-based lexicon model optimized for social media text
- **TextBlob** — Statistical pattern recognition using naive Bayes classification
- **DistilBERT** — Transformer-based deep learning model (distilled from BERT) fine-tuned on SST-2 sentiment dataset
- **RoBERTa** — Robustly optimized transformer architecture trained on Twitter sentiment data by Cardiff NLP

Ensemble Method: Final predictions were determined using majority voting across all models, providing robust and balanced sentiment classification.