

DBL Cork - Presentation 2

1



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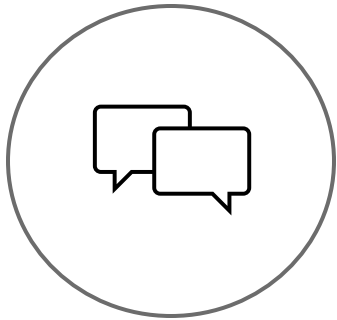
Introduction

Focus:
Sentimental
Analysis for
Lufthansa

Goal:
Understand how
Interactions
Influence a
Customer's
Sentiment

**Competitor
Benchmark:**
Is Lufthansa
Overperforming
or Underperforming

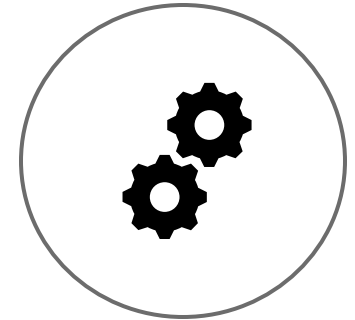
CONTENTS



CONVERSATIONS



SENTIMENT ANALYSIS



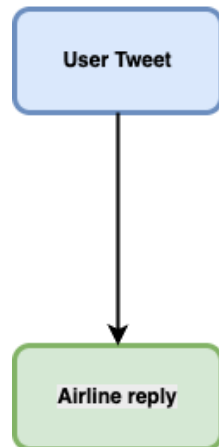
SENTIMENT EVALUATION

Conversations

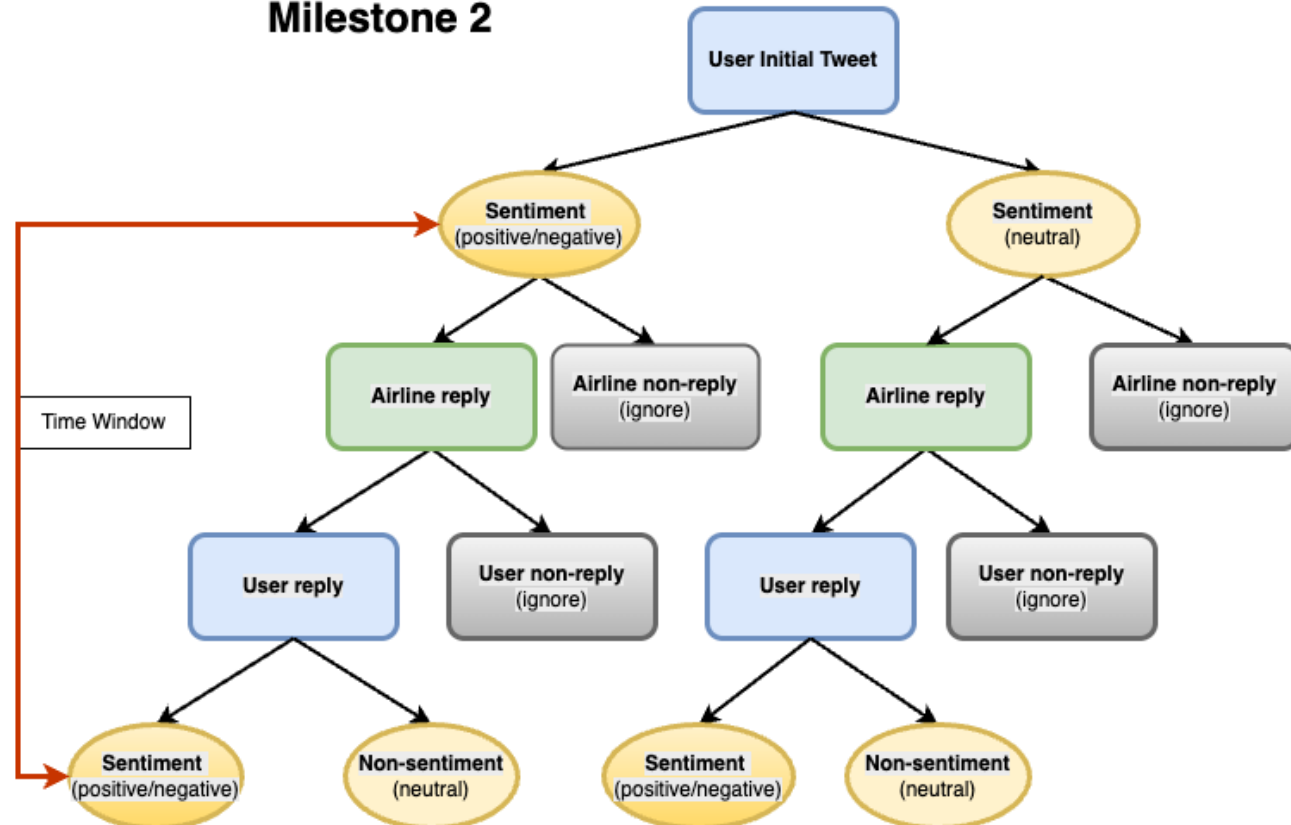
WHAT HAS CHANGED & WHAT IS NEW

Conversation – definition refinement

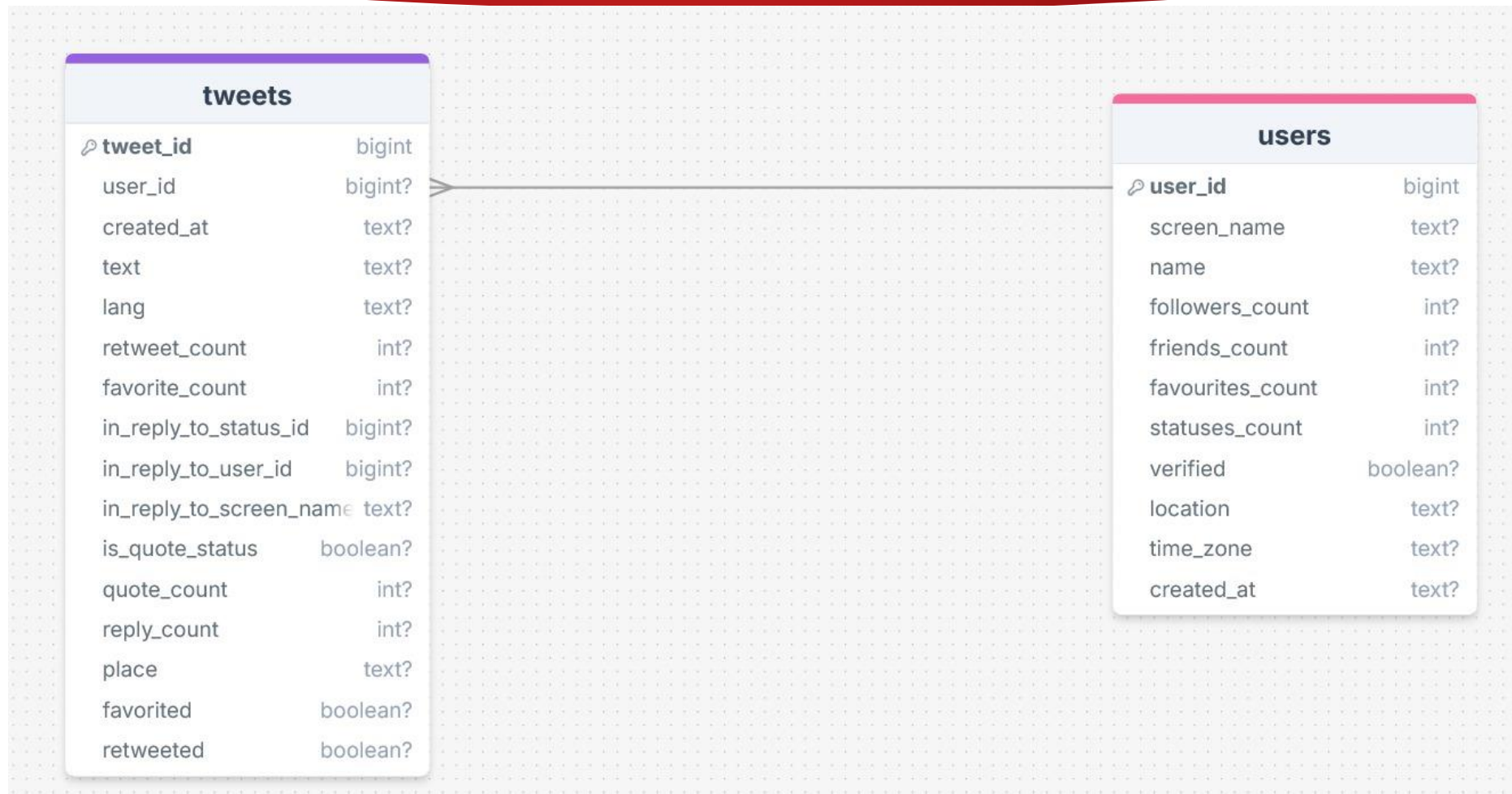
Milestone 1



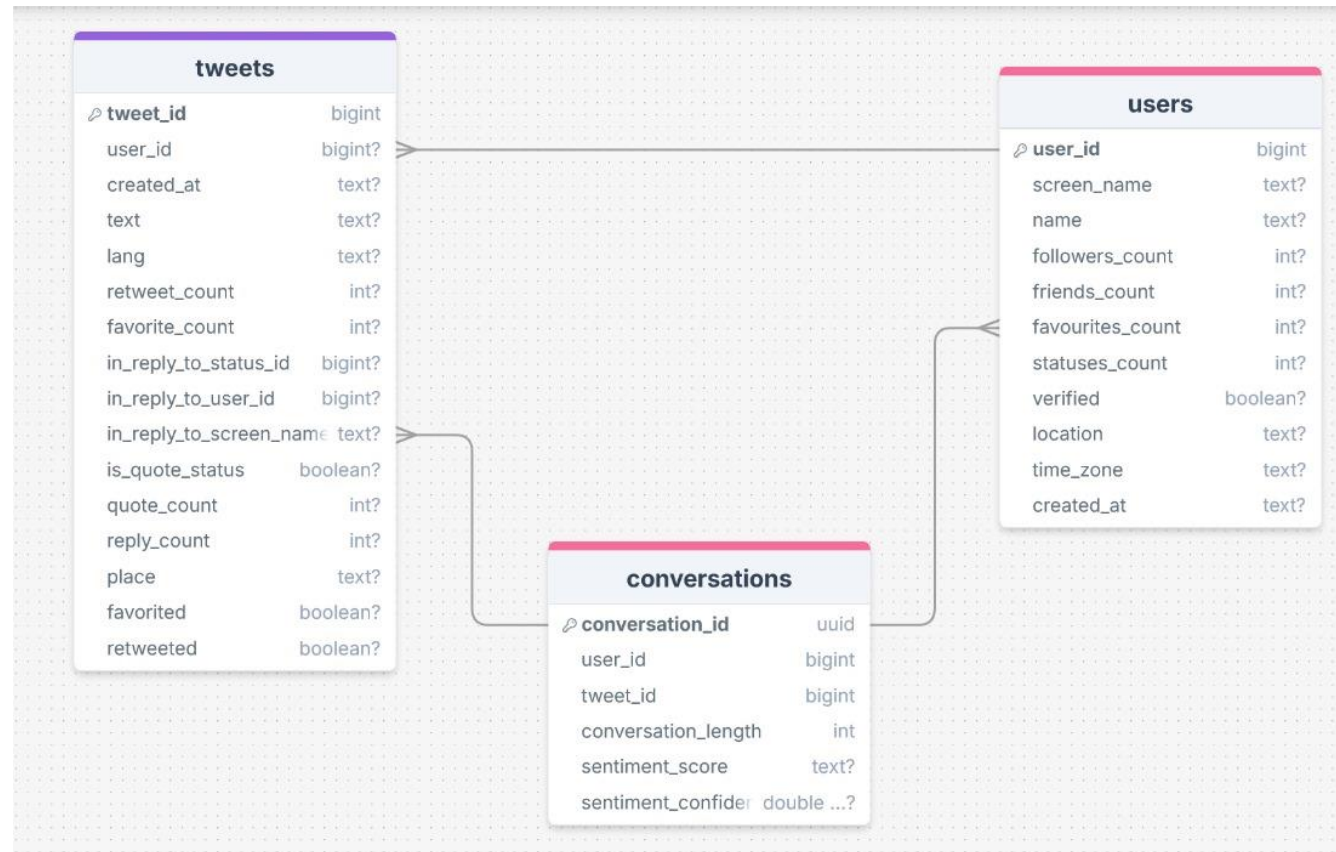
Milestone 2



How its stored: Database - Old



How its stored: Database - New

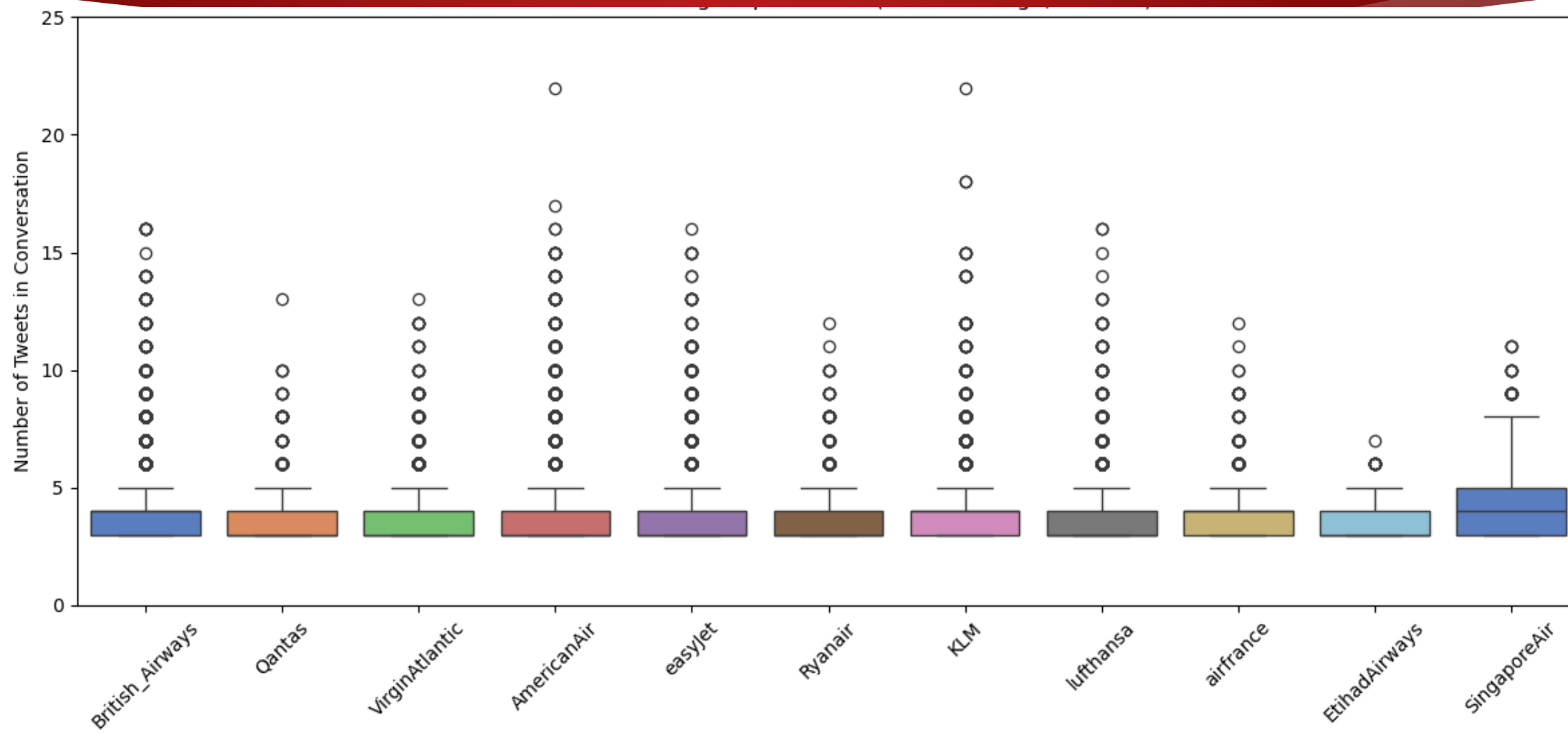


Mean
conversation
length per airline

| Airline | Mean |
|-----------------|----------|
| SingaporeAir | 4.11744 |
| KLM | 4.031275 |
| British Airways | 3.945503 |
| Lufthansa | 3.927701 |
| AirFrance | 3.898973 |
| EasyJet | 3.871400 |
| AmericanAir | 3.849917 |
| VirginAtlantic | 3.661266 |
| Qantas | 3.660830 |
| Ryanair | 3.561219 |
| EihadAirways | 3.518219 |



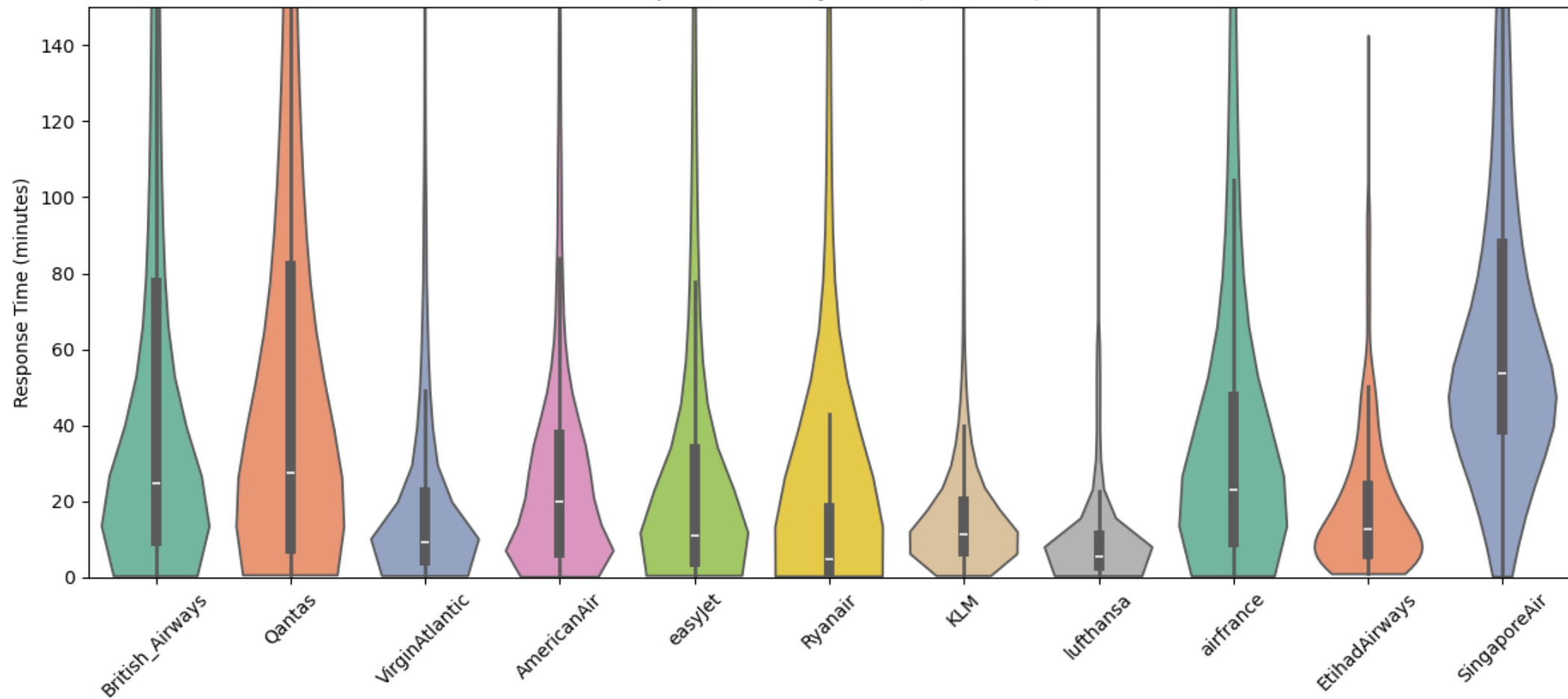
LENGTH OF CONVERSATION PER AIRLINE



Mean response
time per Airline

| Airline | Response Time (Mins) |
|-----------------|----------------------|
| British Airways | 74.13 |
| SingaporeAir | 73.24 |
| Qantas | 69.97 |
| AirFrance | 46.72 |
| Ryanair | 45.20 |
| EasyJet | 43.41 |
| AmericanAir | 31.10 |
| VirginAtlantic | 21.96 |
| KLM | 19.85 |
| EithadAirways | 18.90 |
| Lufthansa | 10.78 |

RESPONSE TIME PER AIRLINE IN MINUTES



Sentiment Analysis

SENTIMENT ANALYSIS - TECHNIQUE

Multilingual

Supports 30+ languages, ideal for international tweets.

-Twitter-Optimized

Trained on real tweets → understands emojis, hashtags, mentions, slang.

-High Accuracy

Outperforms general models like BERT in sentiment classification.

-Lightweight & Fast

Efficient on CPUs and compatible with Apple M1/M2 chips.

WHY WE USE XLM-RoBERTa (CardiffNLP)?

CARDIFFNLP'S XLM RoBERTa model is specifically designed for sentiment analysis on multilingual tweets.

It builds on Facebook AI's XLM-RoBERTa and is fine-tuned using real-world Twitter data.



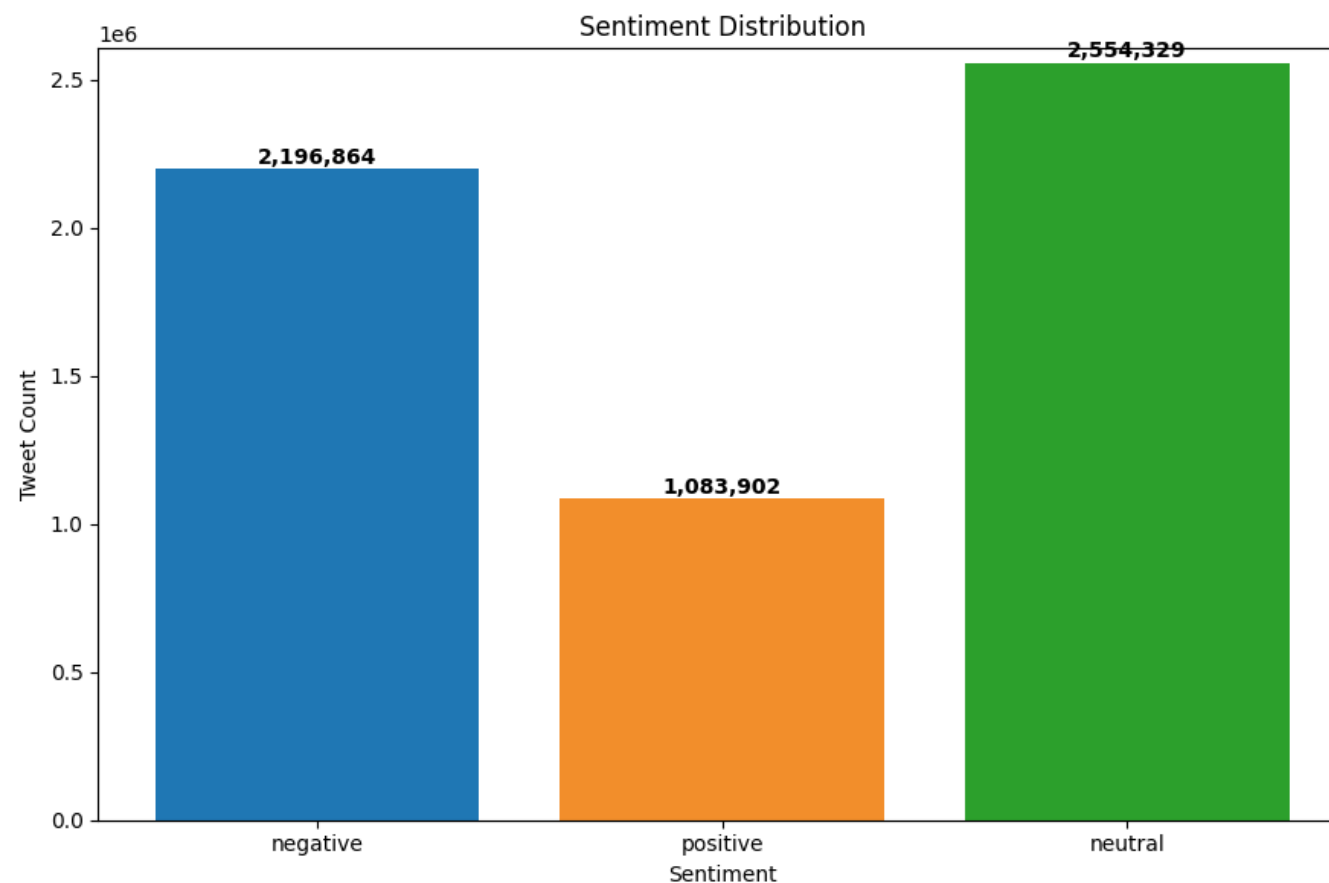
DISTRIBUTION OF THE SENTIMENT

Bar chart indicates that most customer interactions are either neutral in tone or contain complaints, while positive experiences are less frequently expressed.

NEGATIVE – 37,65%

POSITIVE – 18,57%

NEUTRAL – 43,78%

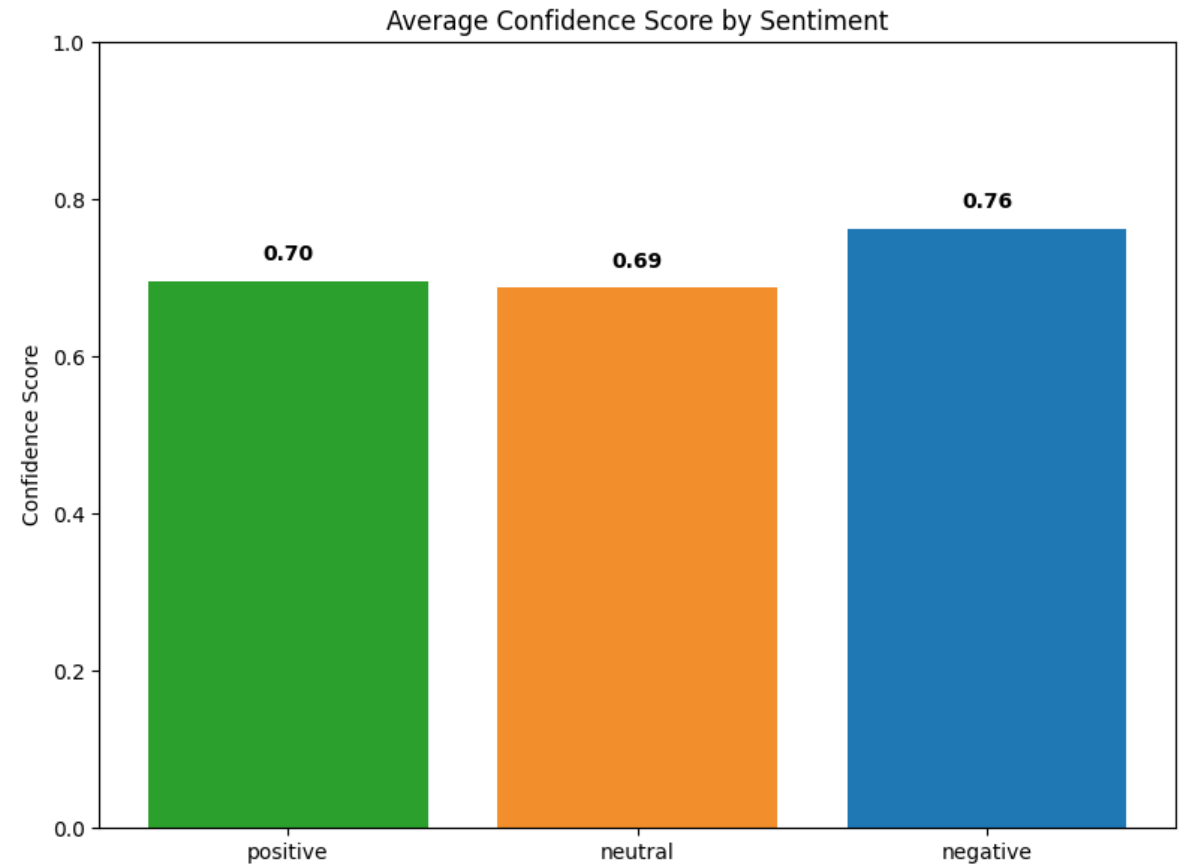


CONFIDENCE LEVEL OF THE SENTIMENT ANALYSIS

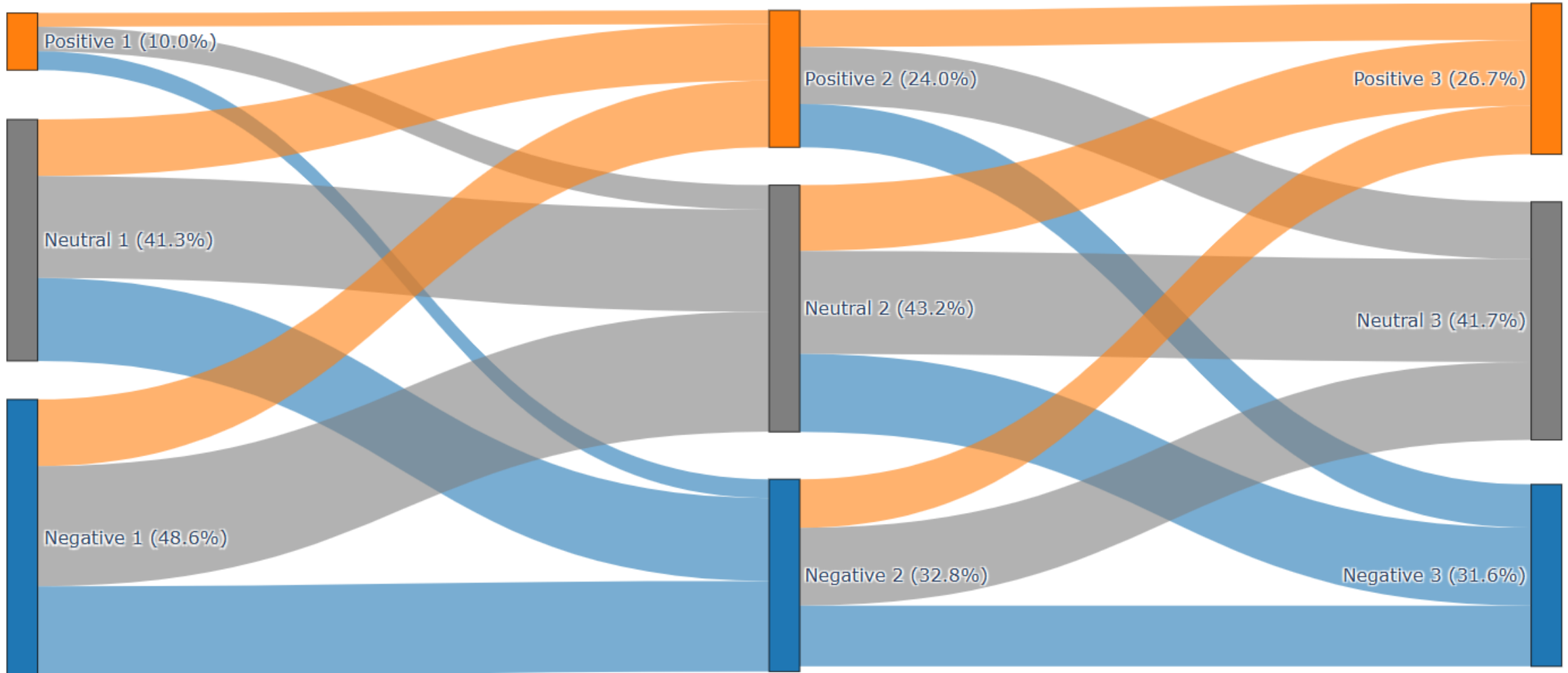
WHY DO WE USE CONFIDENCE SCORES?

Confidence score shows how certain the model is about its sentiment prediction.

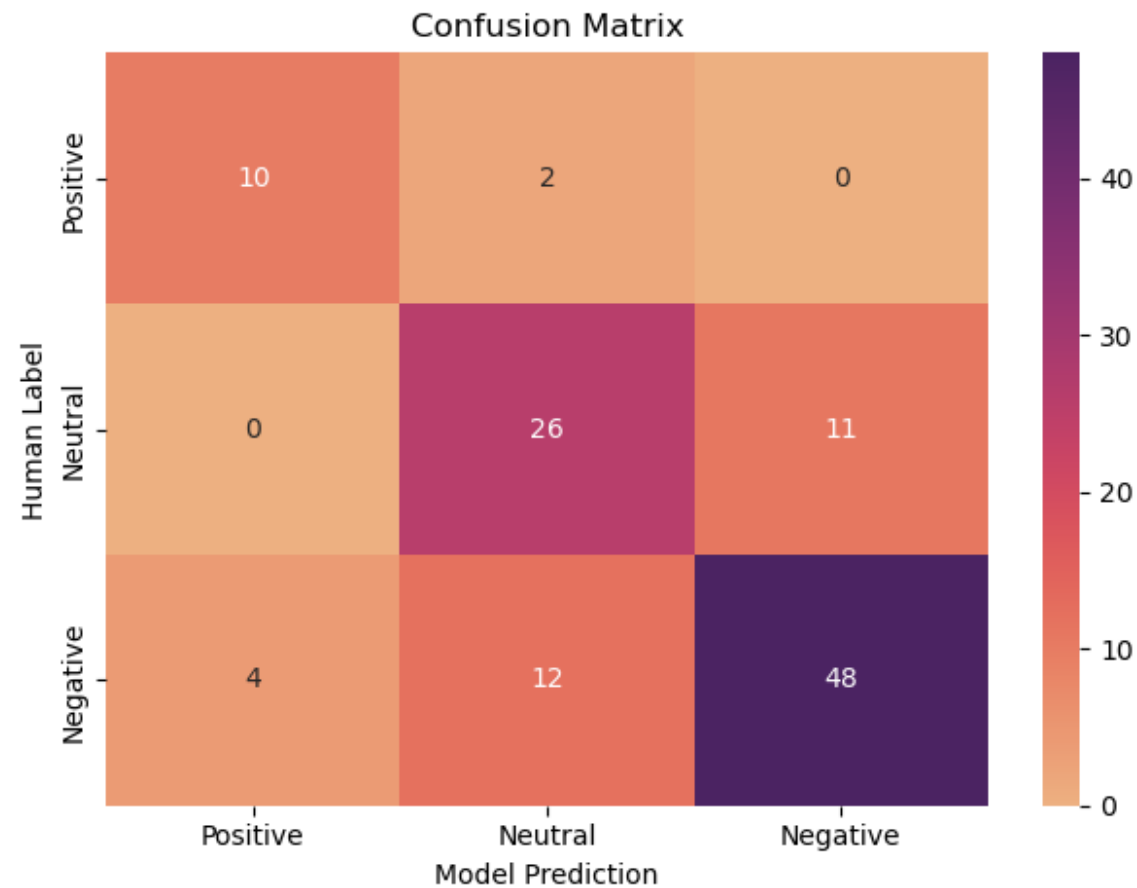
- Helps detect uncertain or borderline tweets
- Useful for filtering out low-confidence predictions
- Allows ranking tweets by prediction reliability
- Supports further analysis (e.g., bias, sensitivity)



Sankey Diagram



SENTIMENT ANALYSIS - EVALUATION

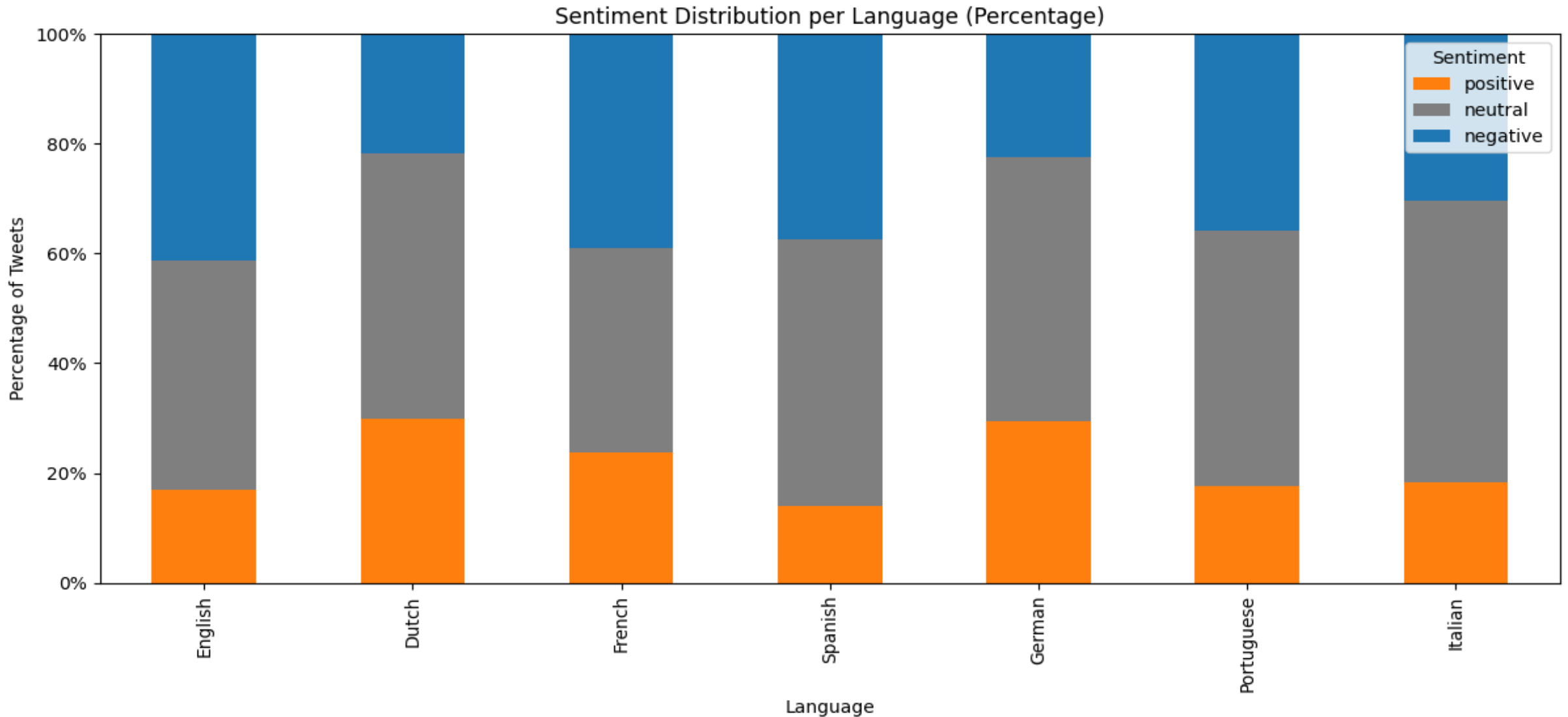


Sentiment Analysis - Evaluation

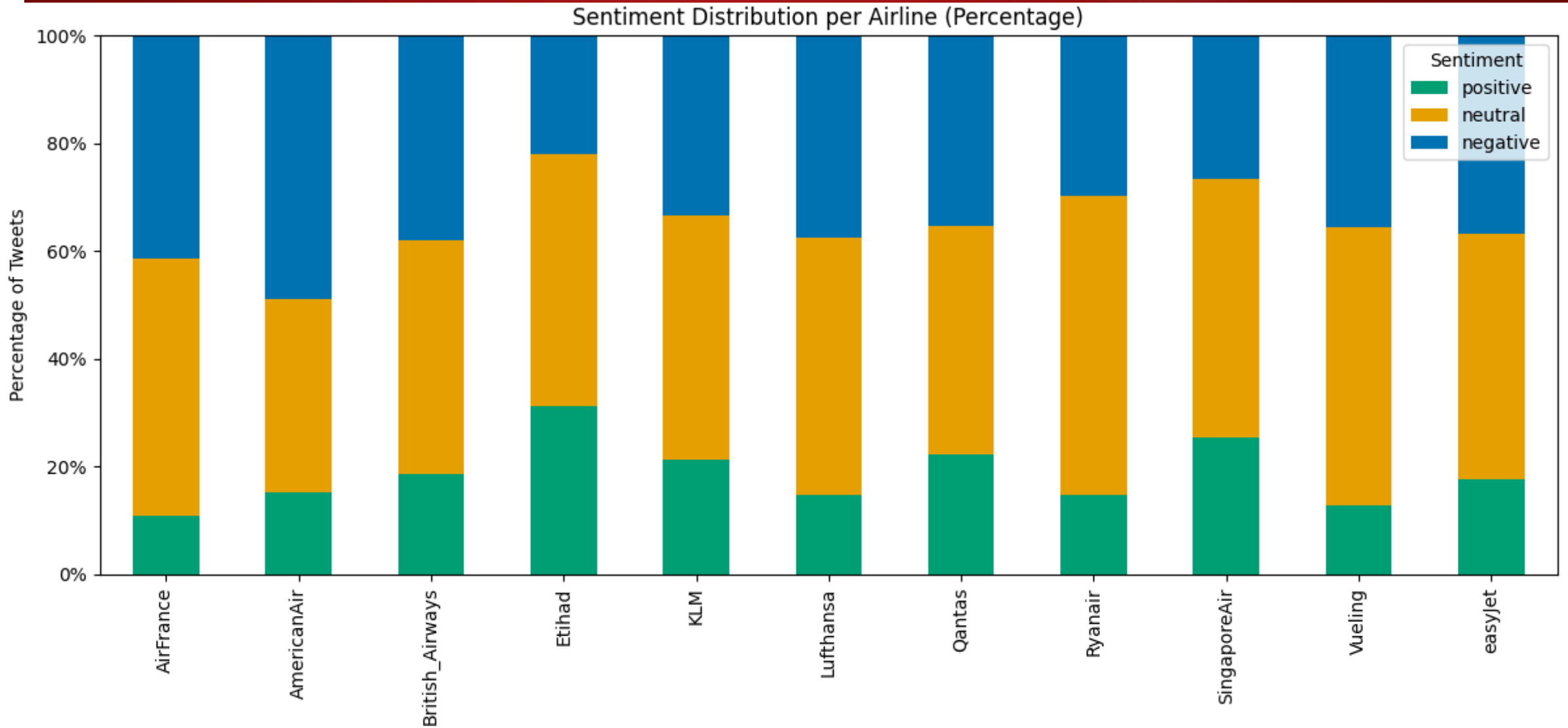
| | Precision | Recall | F1-Score | Support |
|--------------|-----------|--------|----------|---------|
| Negative | 0,81 | 0,75 | 0,78 | 64 |
| Neutral | 0,65 | 0,70 | 0,68 | 37 |
| Positive | 0,71 | 0,83 | 0,77 | 12 |
| Accuracy | - | - | 0,74 | 113 |
| Macro avg | 0,73 | 0,76 | 0,74 | 113 |
| Weighted avg | 0,75 | 0,74 | 0,74 | 113 |

Sentiment Level Per language

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Sentiment Level Per Airline



Conclusion

Lufthansa current position:

- shortest response time
- low positive interactions

Lufthansa's next step:

- understand the customers better
- improve their customer service

