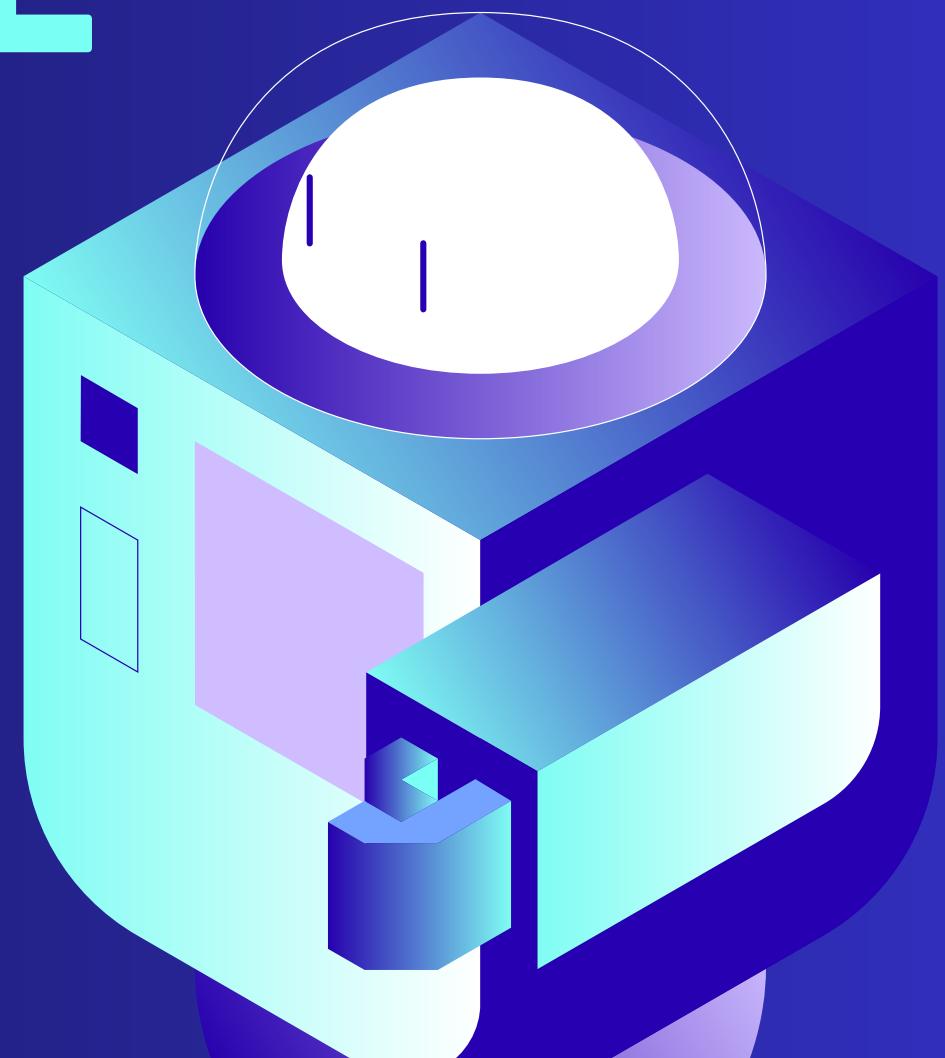


NATURAL LANGUAGE PROCESSING: EMOTION CLASSIFICATION

Raya-Neda, Petar, Vladislav, Mario

GROUP 24



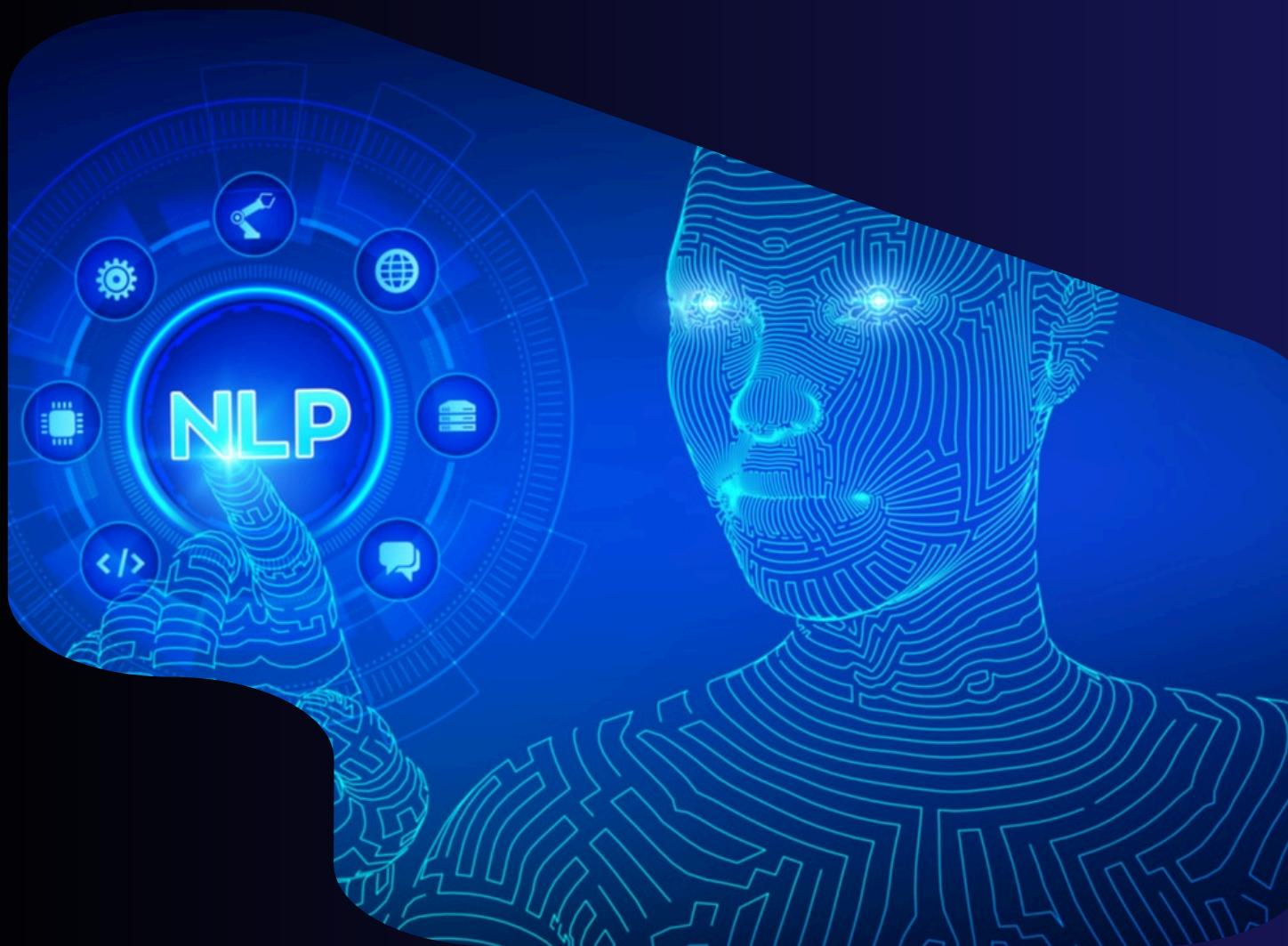
01



AGENDA

- Introduction
 - Client Value
- Dataset
 - Feature Extraction
- Implemented Models
 - Strengths & Weaknesses
- Error Analysis
- Annotations
- Limitations
- Next Steps & Recommendations

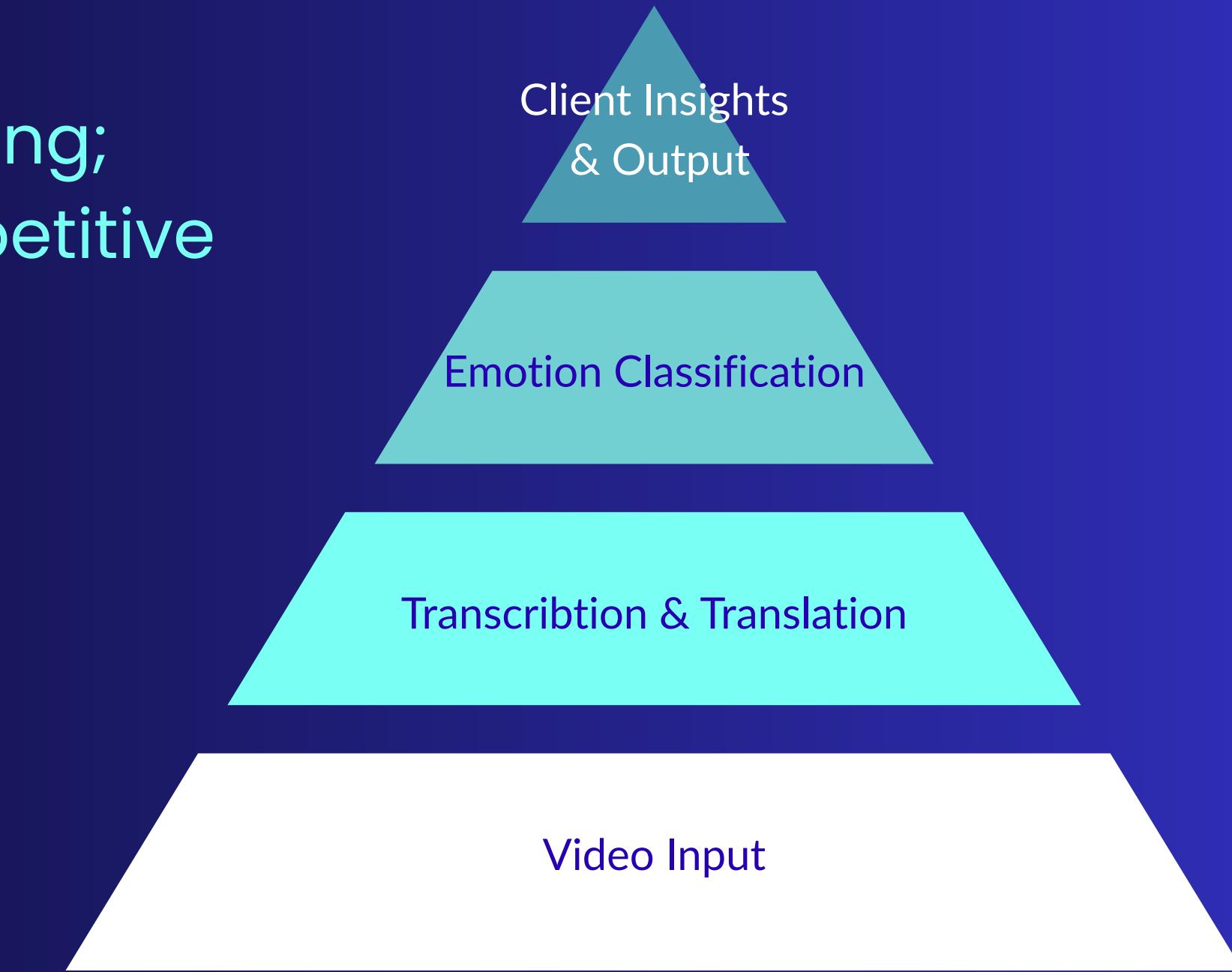
INTRODUCTION



- Problem Definition
 - Emotion Classification in NLP
- Use case
 - Identify emotions expressed in video content to support media creators
- Client
 - CIA – They use AI to analyse every single detail of your film, series, TV-show or video

CLIENT VALUE

- Time & Cost Savings
 - Automates emotional tagging;
 - Reduces human error in repetitive tasks.

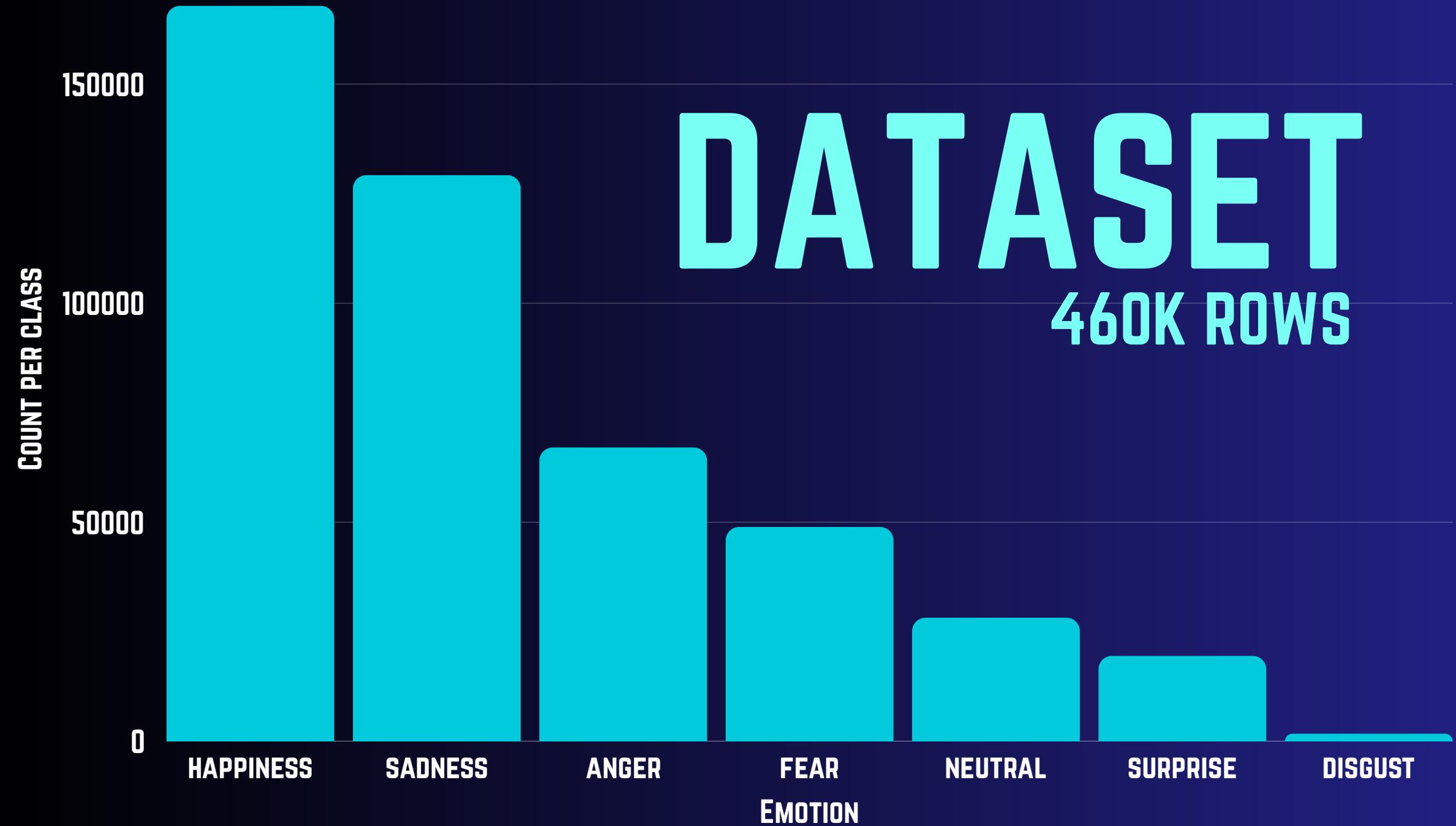


PRE-PROCESSING

Used Datasets

MELD, GoEmotions, CAREER

DATASET
460K ROWS



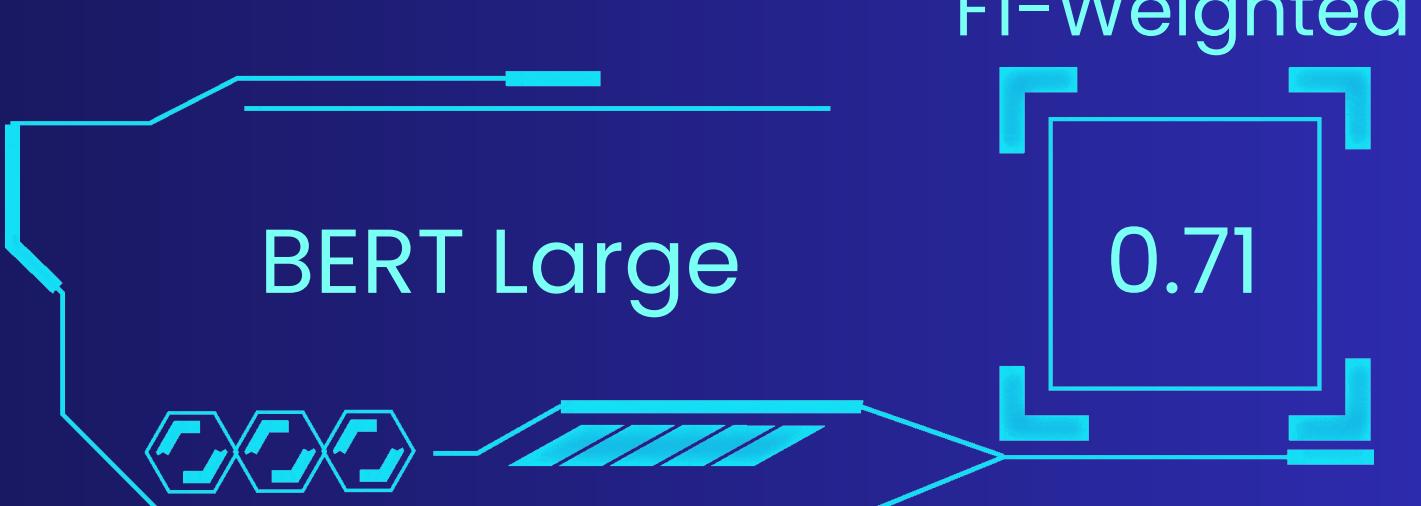
FEATURE EXTRACTION

- POS Tagging
- CountVectorizer & TF-IDF
- Sentiment Scores
- Embeddings



IMPLEMENTED MODELS

APPROACH	WEIGHTED F1-SCORE
Naive Bayes	0.304
Logistic Regression	0.332
RNN	0.32
LSTM	0.31

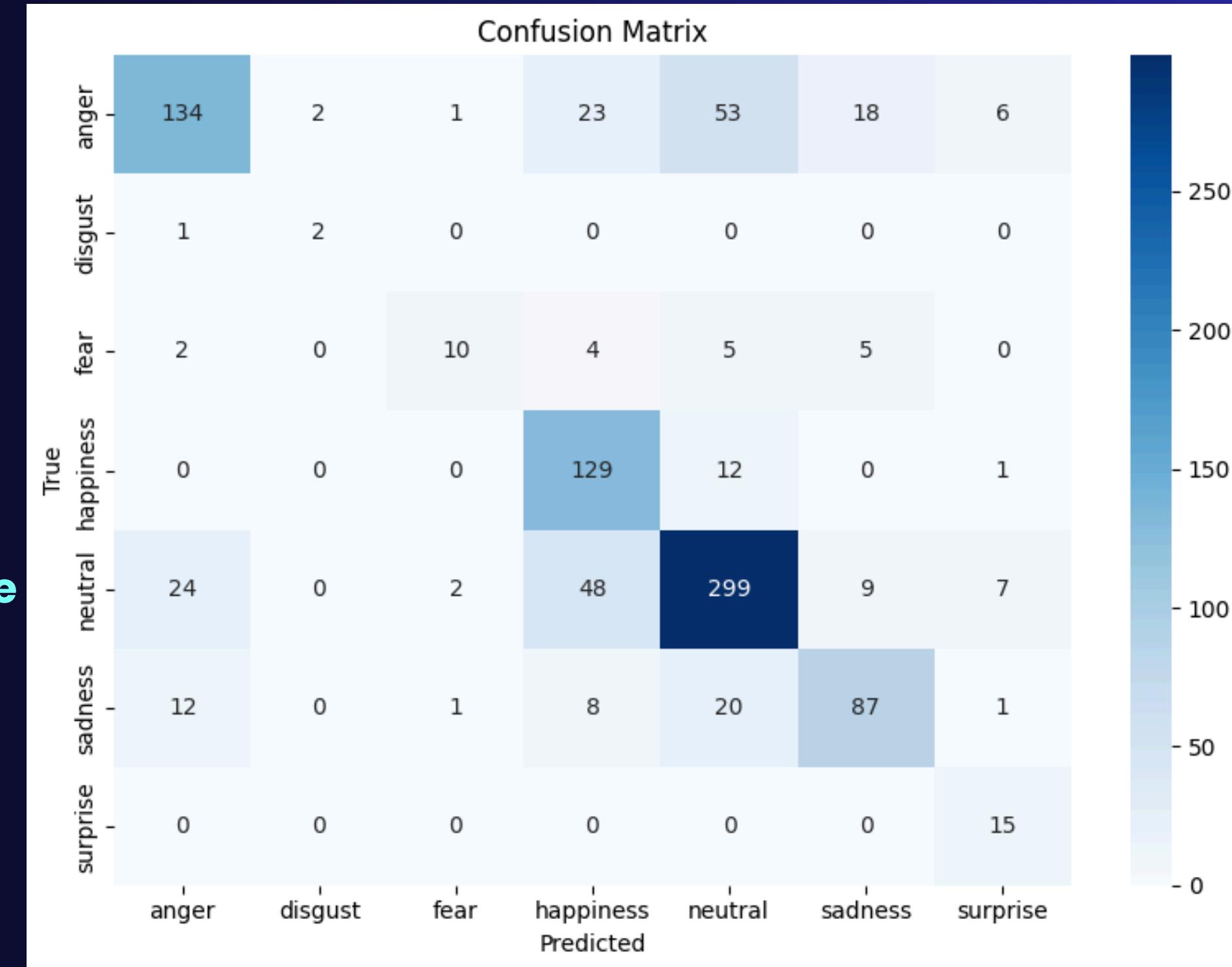


RESULTS

FINAL MODEL

STRENGTHS

- High precision and recall for neutral and happiness
- Anger is recognized **well** with some overlap with **sadness** and **neutral**
- Few false positives for **rare** classes like **disgust** and **surprise**



- Errors between **anger**, **sadness**, and **neutral**
- **Fear** often confused with **neutral** and **happiness**
- Poor performance on **disgust** and **surprise**

WEAKNESSES

emotions	precision	recall	f1-score
anger	0.77	0.57	0.65
disgust	0.5	0.67	0.57
fear	0.71	0.38	0.5
happiness	0.60	0.91	0.73
neutral	0.77	0.77	0.77
sadness	0.73	0.67	0.70
surprise	0.5	1	0.67
macro avg	0.66	0.71	0.66
weighted avg	0.73	0.72	0.71

ERROR ANALYSIS

ERROR ANALYSIS

- Label Issues
 - Emotional sentences mislabelled as 'neutral'
 - Flat sentences labeled as 'happiness'
 - Translation/Transcription Errors
 - Lost words & incomplete sentences
 - Unclear phrases due to poor transcription
 - CIA Test Set Fixes
 - Corrected many wrong labels & translations
 - Improved evaluation reliability

EXAMPLES

Don't worry. CIA: neutral Actual: happiness

We immediately go home. CIA: happiness, Actual: neutral

For my daughter. CIA: happiness, Actual: neutral

We were better with my father. CIA: neutral, Actual: sadness (nostalgia)

Like Sashka's dog. CIA: happiness, Actual: neutral

But there is no way to do because of CIA: anger, Actual: sadness (incomplete sentence)

Lilyana has been around the neighborhood several times CIA: happiness, Actual: neutral

that she

has a major responsibility
on your life and love.

Incorrect Transcription (One sentence divided as 3 distinct)

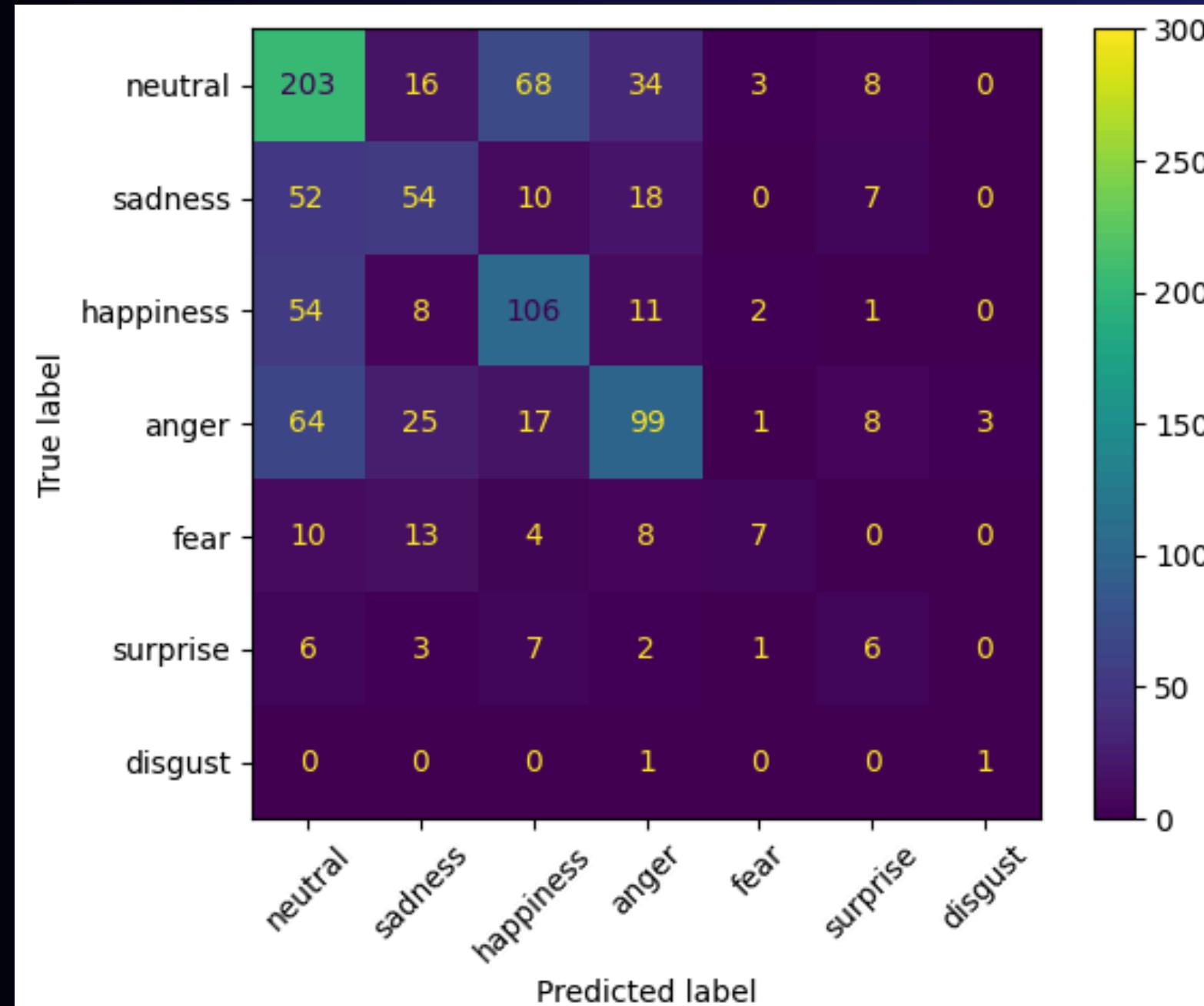
The idea of Äcä, -â€¹Äcä, -â€¹the box sounds very good to me.

Example of Incorrect Translation 14

I am lumpy that Velina

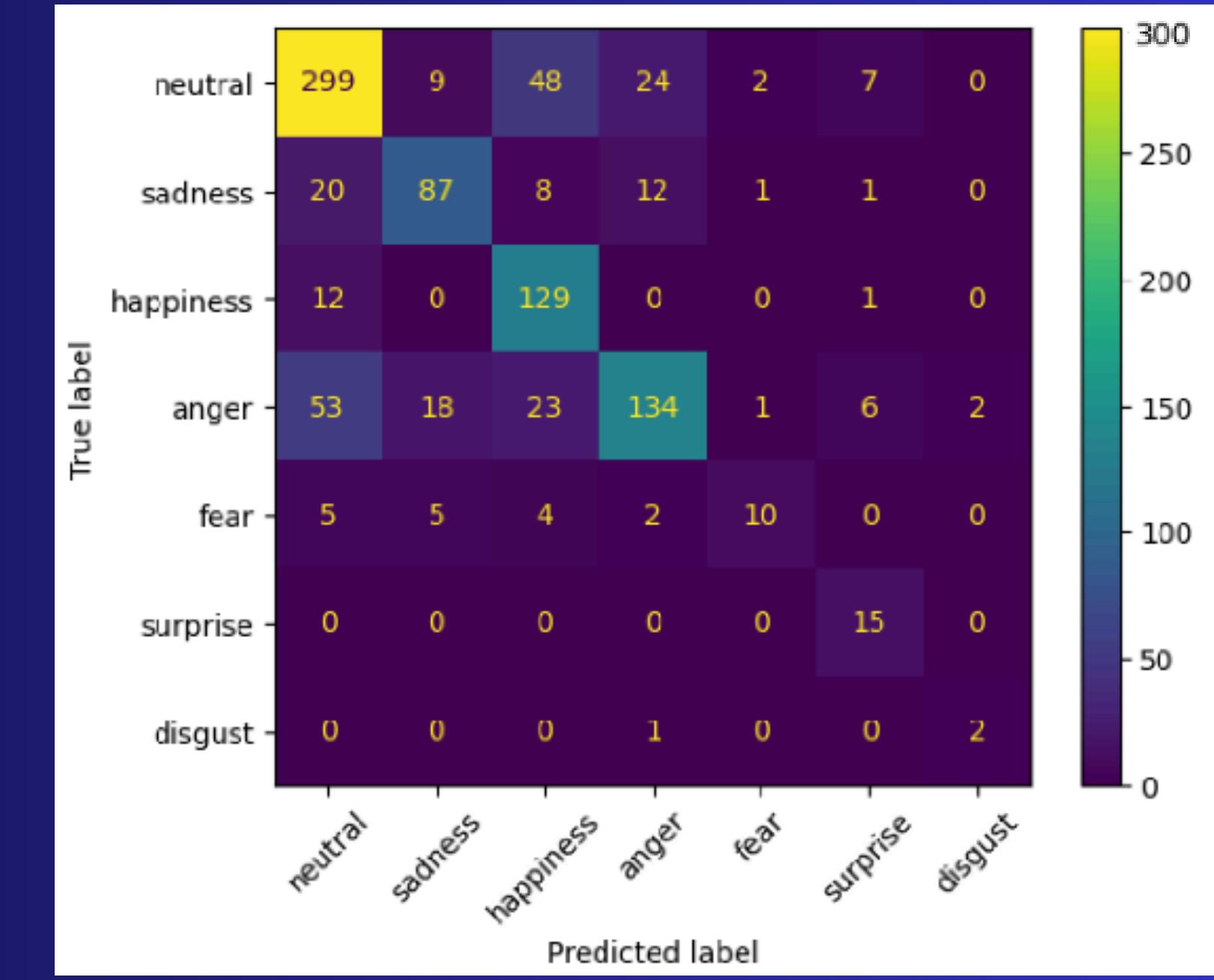
Example of Incorrect Transcription

ANNOTATIONS



Before

F1-Weighted = 0.499



After

F1-Weighted = 0.714

LIMITATIONS

- Data Imbalance
 - Incomplete transcription
 - Translation Errors
 - Context Loss in Text-only Analysis
- 

NEXT STEPS & RECOMMENDATIONS

- Improve data quality
- Collect more diverse training data
- Implement the whole pipeline into an application



THANK YOU

FOR YOUR ATTENTION

Presented By: Olivia Wilson

Raya-Neda, Petar, Vladislav, Mario

