



Team MoodSwing Lyrical Mood Classification

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01

Goal & Motivation





What

●

**Accurately classifying
song lyrics into moods
(happy, sad, angry,
relaxed)**

Why

●

**Aid in understanding the
emotional impact of songs
on listeners and improve
music recommendations**

How

●

?



The slide features a light green background with decorative illustrations in the corners. The top-left corner shows a green and yellow striped conical object, possibly a hat or a musical instrument, with two green sticks. The top-right corner shows a green leafy plant with a brown and black object. The bottom-left corner shows a green leafy plant with a red object. The bottom-right corner shows a brown and yellow striped drum with two brown sticks, next to a green leafy plant with a red object.

02 Data & Data Processing

MoodyLyricQ Dataset

- ❑ Curated **collection of song titles, associated artist names** spanning across various genres.
- ❑ Individually **labeled with emotional tags** for every song – Happy, Sad, Angry, and Relaxed. [Each song has a **single label**.]
- ❑ The following dataset **does not contain song lyrics** due to copyright issues.

Mood	Number of Song Lyrics Available
Happy	500
Relaxed	423
Sad	415
Angry	383
Total	1721

Obtaining Lyrics

❑ **Genius API Integration:**

- ❑ Using the Genius API's search feature for fetching lyrics using a combination of artist's name and song name.

❑ **File Organization:**

- ❑ Retrieved lyrics and metadata are stored in individual .txt, .json files, respectively.

Mood	Total Lyrics Scraped	Subset used for training
Happy	500	45
Relaxed	423	45
Sad	415	45
Angry	383	45
Total	1721	180

Preprocessing

Remove extra lines and metadata

Remove stop words

Apply stemming and lemmatization

Manually verify 180 samples

Create a perfectly class-balanced dataset

Data Sample

	Index	Artist	Title	Mood	Filename	Lyrics
0	ML100	Bad Company	Can't Get Enough	happy	ML100_Cant Get Enough.txt	[Intro]\n\n[Verse 1]\nWell, I take whatever I ...
1	ML1003	Los Lobos	Come On Let's Go	happy	ML1003_Come On Lets Go.txt	[Verse 1]\nWell, come on let's go, let's go, l...
49	ML1000	The Police	So Lonely	sad	ML1000_So Lonely.txt	\n[Verse 1]\nWell, someone told me yesterday\n...
50	ML102	Billy Squier	In The Dark	sad	ML102_In The Dark.txt	\nLife isn't easy from the singular side\nDown...
79	ML1010	The Meters	Fire On The Bayou	angry	ML1010_Fire On The Bayou.txt	\nFire on the bayou\nFire on the bayou\n\nDown...
80	ML1025	Metric	On The Sly	angry	ML1025_On The Sly.txt	\n[Verse 1]\nWipe it up you're capable\nOf put...
127	ML1	Usher	There Goes My Baby	relaxed	ML1_There Goes My Baby.txt	\n[Intro]\nYeah... right...\nUsher baby... oka...
128	ML10	Beenie Man	Slam	relaxed	ML10_Slam.txt	\nWell, this is like\nA menace to the society\n...

Data frame

ML1032_Because Of You.txt

[Verse]
Because of you, there's a song in my heart
Because of you, my romance had its start
Because of you, the sun will shine
The moon and stars will say you're mine
Forever, and never to part

[Chorus]
I only live for your love and your kiss
It's paradise to be near you like this
Because of you, my life is now worthwhile
And I can smile
because of you

[Instrumental Break]

[Chorus]
I only live for your love and your kiss
It's paradise to be near you like this
Because of you, my life is now worthwhile
And I can smile
because of you

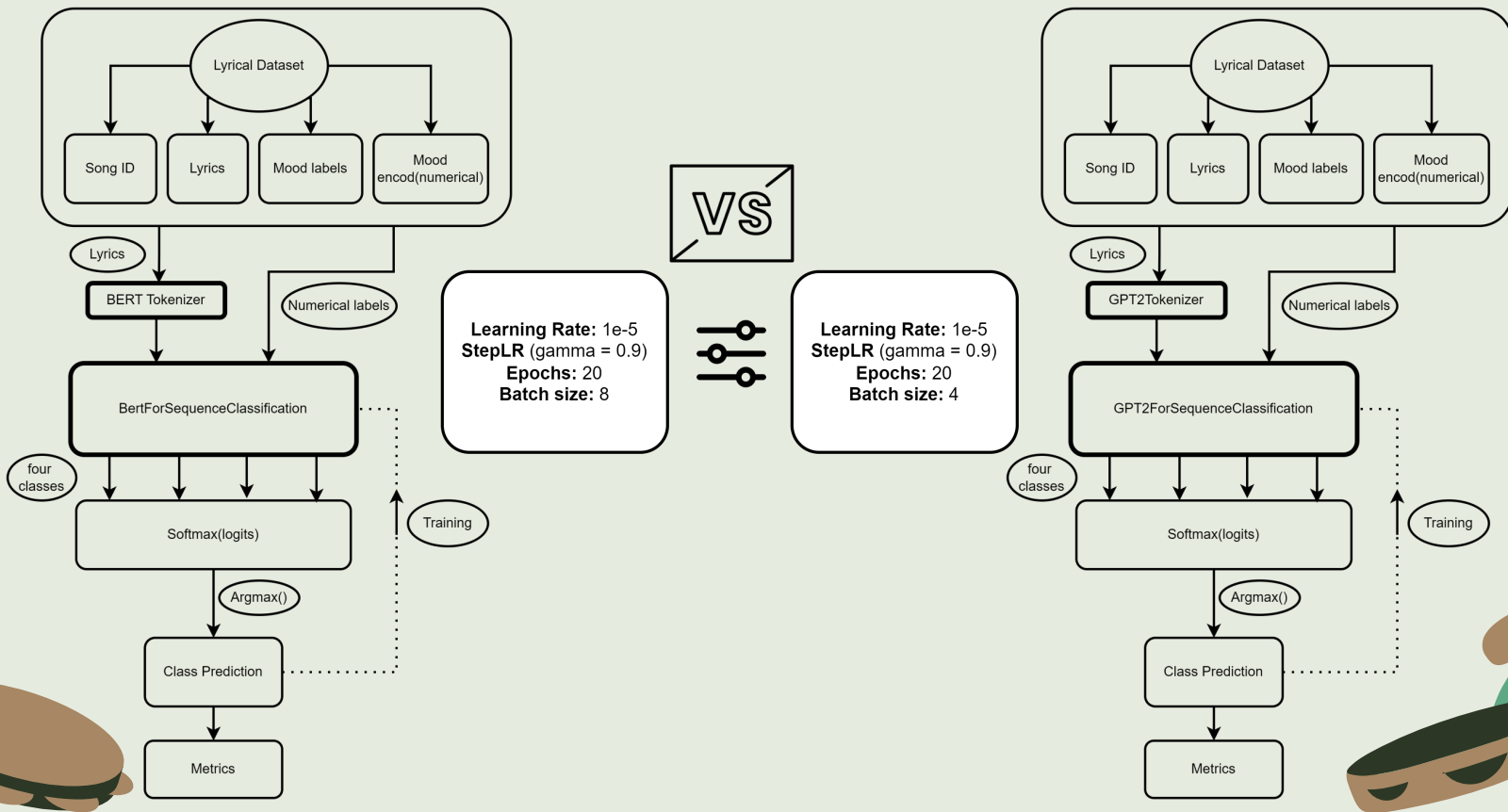
Lyrics



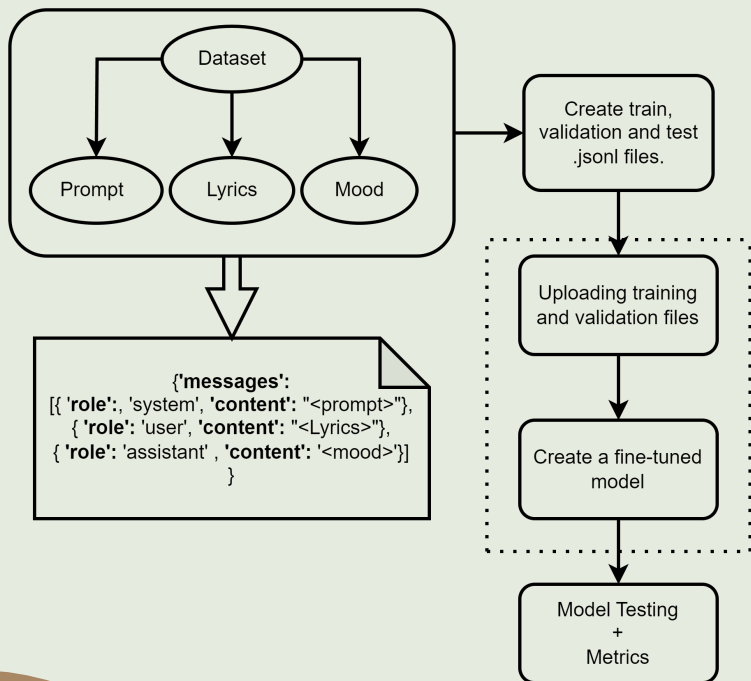
03 Models & Software



BERT vs GPT2



GPT3.5 Model



PROMPT|

1

"You are a chatbot that, when prompted with song lyrics, predicts one of the emotions ('Happy', 'Sad', 'Angry', or 'Relaxed') without providing any explanation. Reply with only the emotion name."

You do not retain any previous information regarding the lyrics given to you. You specialize in analyzing the given song lyrics and predicting the emotion of the song."

VS

2

"You are a chatbot that, when prompted with song lyrics, predicts one of the emotions ('Happy', 'Sad', 'Angry', or 'Relaxed') without providing any explanation. Reply with only the emotion name. "



04 Results

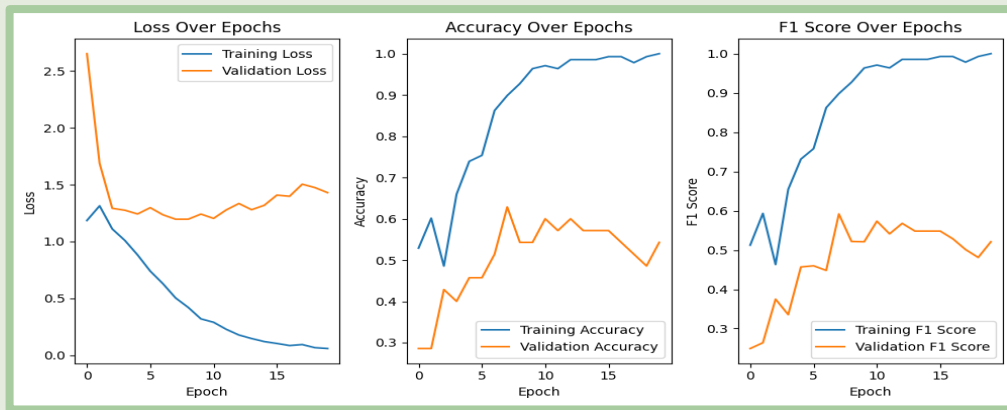


Quantitative Results



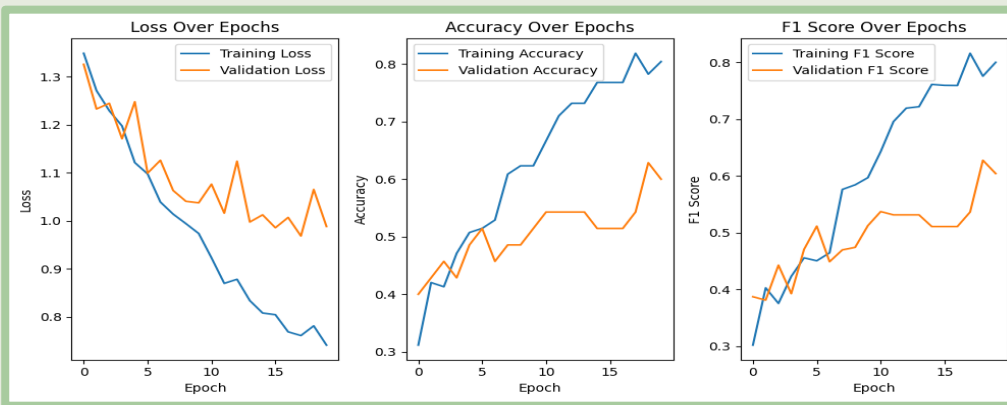
Training & Validation Curves

GPT2



VS

BERT



Baseline Model Performance

Classification Report - Validation:

	precision	recall	f1-score	support
happy	0.71	0.71	0.71	14
sad	0.50	0.60	0.55	5
angry	0.71	0.56	0.63	9
relaxed	0.38	0.43	0.40	7
accuracy			0.60	35
macro avg	0.58	0.57	0.57	35
weighted avg	0.62	0.60	0.60	35

BERT

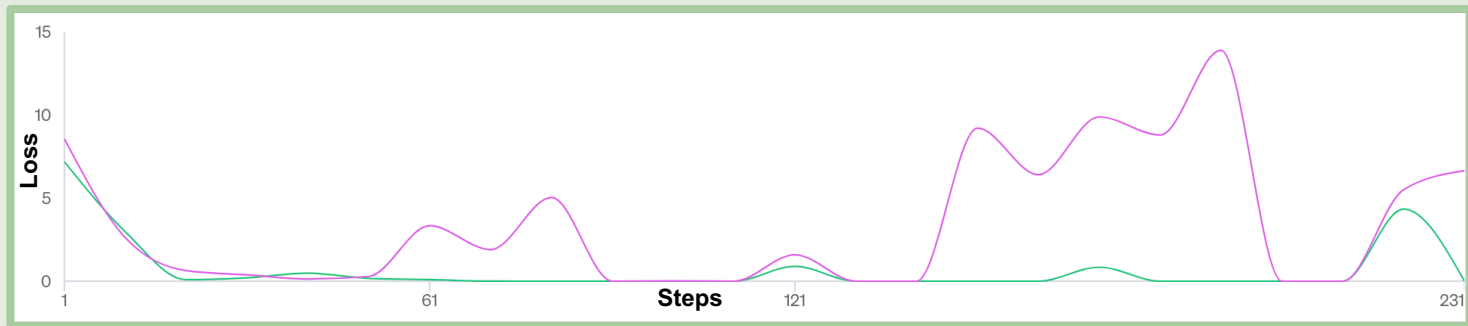
VS

GPT2

Classification Report - Validation:

	precision	recall	f1-score	support
happy	0.67	0.86	0.75	14
sad	0.25	0.40	0.31	5
angry	0.67	0.44	0.53	9
relaxed	0.33	0.14	0.20	7
accuracy			0.54	35
macro avg	0.48	0.46	0.45	35
weighted avg	0.54	0.54	0.52	35

OpenAI GPT3.5 Model Performance



Testing Sample of 80 Lyrics: Classification Report

	precision	recall	f1-score	support
happy	0.78	0.58	0.67	24
sad	0.54	0.58	0.56	12
angry	0.74	0.74	0.74	19
relaxed	0.52	0.67	0.59	18
accuracy			0.64	73
macro avg	0.64	0.64	0.64	73
weighted avg	0.66	0.64	0.65	73

- ❑ Strong performance in identifying 'Angry', 'Happy' emotions.
- ❑ 'Relaxed' show decent precision but lower recall, with occasional misses.
- ❑ 'Sad' faces challenges with lower precision, F1-score, indicating difficulty in classification.

Qualitative Analysis





Correct & Misclassified Examples

[Verse]
Because of you, there's a song in my heart
Because of you, my romance had its start
Because of you, the sun will shine
The moon and stars will say you're mine
Forever, and never to part

[Chorus]
I only live for your love and your kiss
It's paradise to be near you like this
Because of you, my life is now worthwhile
And I can smile
because of you

[Instrumental Break]

[Chorus]
I only live for your love and your kiss
It's paradise to be near you like this
Because of you, my life is now worthwhile
And I can smile
because of you

Actual Label:
Happy
Predicted Label:
Relaxed
Explanation:
The mention of a
"song in my heart,"
"paradise," and the
overall soothing
tone could
contribute to a
sense of relaxation

Actual Label:
Angry
Predicted Label:
Angry
Explanation:
The presence of
phrases such as
"evil minds,"
"destruction," and
expressions of
contempt for war
elicits feelings of
anger.

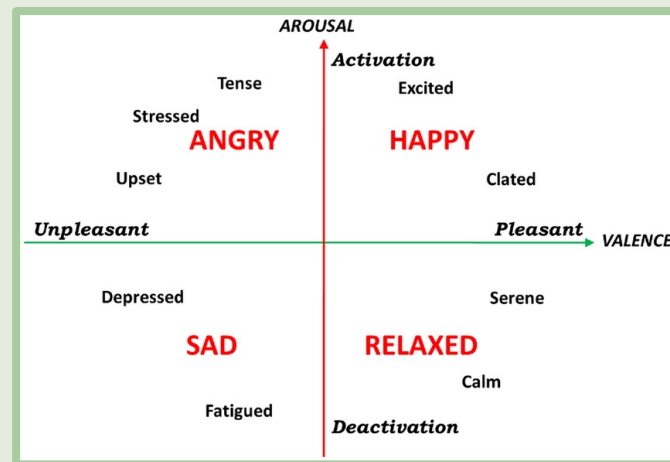
[Verse 1]
Generals gathered in their masses
Just like witches at black masses
Evil minds that plot destruction
Sorcerers of death's construction
In the fields, the bodies burning
As the war machine keeps turning
Death and hatred to mankind
Poisoning their brainwashed minds
Oh, Lord, yeah

[Bridge]
Politicians hide themselves away
They only started the war
Why should they go out to fight?
They leave that all to the poor, yeah
Time will tell on their power minds
Making war just for fun
Treating people just like pawns in
chess
Wait till their judgment day comes,
yeah



General Trends Observed

Emotion	Highest Misclassified label
Happy	Relaxed
Sad	Angry
Angry	Sad
Relaxed	Happy





05

Discussion and Learnings



Interesting Findings

01



BERT performs better than GPT2- due to its bi-direction context learning

02



Minimal pre-processing gives better performance of models

03



Class 'Relaxed' consistently tends to be misclassified as 'Happy'

04



The model is inclining towards valence more than arousal



Thanks!

Do you have any questions?

