

Emotion Patterns in Music Playlists

Sara Giammusso¹² Mario Guerriero¹²

¹MSc student in Data Science Department, EURECOM, Télécom ParisTech, France

²MSc student in Department of Control and Computer Engineering, Politecnico di Torino, Italy

Third Project meeting

Table of Contents

- 1 Introduction
- 2 LyricsKMeans: Model Description
- 3 LyricsKMeans: Model Evaluation
- 4 LyricsNN: Model Description
- 5 LyricsNN: Model Evaluation
- 6 Conclusion
- 7 References

Table of Contents

- 1 Introduction
- 2 LyricsKMeans: Model Description
- 3 LyricsKMeans: Model Evaluation
- 4 LyricsNN: Model Description
- 5 LyricsNN: Model Evaluation
- 6 Conclusion
- 7 References

Introduction

Previously On Sara&Mario Project...

We analyzed already existent systems and we came up with some decisions about what to do.

Next steps:

- Start building some simple classifiers
- Evaluate those simple classifiers
- See how we could improve it

Table of Contents

- 1 Introduction
- 2 LyricsKMeans: Model Description**
- 3 LyricsKMeans: Model Evaluation
- 4 LyricsNN: Model Description
- 5 LyricsNN: Model Evaluation
- 6 Conclusion
- 7 References

The Idea

- Based on the same idea of the Nearest Neighbour
- Find the "closest" song and get its label
- We are not actually comparing lyrics to each other
- We compare lyrics we want to classify to 4 points
 - Each point is supposed to be representative of an emotion

Why 4 points?

- Simplicity
- Speed
- We just wanted to get some quick insights

How it works

- Evaluate the word vector norm for each lyrics
- Group (compute average) lyrics by emotion
- Given a lyric find the closest representative point

Table of Contents

- 1 Introduction
- 2 LyricsKMeans: Model Description
- 3 LyricsKMeans: Model Evaluation**
- 4 LyricsNN: Model Description
- 5 LyricsNN: Model Evaluation
- 6 Conclusion
- 7 References

1-fold validation

We used the simplest validation technique: 1-fold validation

- 90% for training
- 10% for validation

Accuracy

- Not very impressive
- Around 28% accuracy



But I'll keep trying

Table of Contents

- 1 Introduction
- 2 LyricsKMeans: Model Description
- 3 LyricsKMeans: Model Evaluation
- 4 LyricsNN: Model Description**
- 5 LyricsNN: Model Evaluation
- 6 Conclusion
- 7 References

The Idea

DISCLAIMER

This model is slower than the previous one

- A more general version of the LyricsAverageNN classifier
- Based on the same idea of the k-Nearest Neighbour (again)
- Find the "closest" song and get its label
- Now we are comparing lyrics to each other

How it works

- Evaluate the word vector norm for each lyrics
- Do NOT group anything
- Given a lyric find the closest lyrics and get their majority label

Table of Contents

- 1 Introduction
- 2 LyricsKMeans: Model Description
- 3 LyricsKMeans: Model Evaluation
- 4 LyricsNN: Model Description
- 5 LyricsNN: Model Evaluation**
- 6 Conclusion
- 7 References

1-fold validation

We used the same validation technique we used for the previous model

- 90% for training
- 10% for validation

Accuracy

- Not very impressive
- 33% accuracy ($k=1$)
- 30% accuracy ($k=3$)
- 29% accuracy ($k=5$)



But I'll keep trying

Table of Contents

- 1 Introduction
- 2 LyricsKMeans: Model Description
- 3 LyricsKMeans: Model Evaluation
- 4 LyricsNN: Model Description
- 5 LyricsNN: Model Evaluation
- 6 Conclusion**
- 7 References

What's next?

- We didn't use Fasttext word vectors for the sake of time...
- We need better feature engineering
- Move to better classifiers (it didn't make sense to use them for now)
- Find out some better pre-processing techniques
- Implement better validation technique (cross-validation!!!)

Table of Contents

- 1 Introduction
- 2 LyricsKMeans: Model Description
- 3 LyricsKMeans: Model Evaluation
- 4 LyricsNN: Model Description
- 5 LyricsNN: Model Evaluation
- 6 Conclusion
- 7 References**

References I

- [1] spaCy: Industrial-Strength Natural Language Processing
<https://spacy.io/>
- [2] Intro to NLP with SpaCy
<https://nicschrading.com/project/Intro-to-NLP-with-spaCy/>