

Emotion Patterns in Music Playlists

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Third Project meeting

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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

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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

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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

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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

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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 ($k=5$)

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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!!!)

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