

### Introduction to Audio Content Analysis

module 13: mood recognition

alexander lerch



 overview
 intro
 mood
 models
 regression
 results
 summa

 ●
 ○
 ○
 ○
 ○
 ○
 ○
 ○

### introduction overview

Georgia Center for Music Tech Market Technology

#### corresponding textbook section

#### chapter 13

#### lecture content

- · introduction to emotion and mood
- models for mood
- linear regression

#### learning objectives

- describe Russel's arousal-valence plane
- discuss commonalities and differences between mood recognition and genre classification
- implement linear regression in Matlab



module 13: mood recognition 1 / 9

 overview
 intro
 mood
 models
 regression
 results
 summar

 ●
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○
 ○

### introduction overview



#### corresponding textbook section

#### chapter 13

#### lecture content

- introduction to emotion and mood
- models for mood
- linear regression

#### learning objectives

- describe Russel's arousal-valence plane
- discuss commonalities and differences between mood recognition and genre classification
- implement linear regression in Matlab



module 13: mood recognition 1 / 9

# mood recognition introduction



- objective:identify mood/emotion of a song
- terminology:
  - Music Mood Recognition and Music Emotion Recognition usually used synonymously
- processing steps (similar to genre and similarity tasks)
  - extract features
  - classify (possibly regression)

module 13: mood recognition 2 / 9

# mood recognition introduction



- **objective**:identify mood/emotion of a song
- **■** terminology:
  - Music Mood Recognition and Music Emotion Recognition usually used synonymously
- processing steps (similar to genre and similarity tasks)
  - extract features
  - classify (possibly regression)

# mood recognition introduction



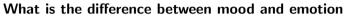
- **objective**:identify mood/emotion of a song
- terminology:
  - Music Mood Recognition and Music Emotion Recognition usually used synonymously
- processing steps (similar to genre and similarity tasks)
  - extract features
  - classify (possibly regression)

Georgia Center for Music Tech || Technology

What is the difference between mood and emotion







many definitions out there but general consensus on

- emotion:
  - temporary, evanescent
  - (directly) related to external stimuli
- mood:
  - longer term, stable
  - diffuse affect state





#### **■** ground truth data

- verbalization of emotions/moods usually misleading
- not easily *quantifiable*/categorizable
- changing over time?

### research questions

- are basic emotions (happiness, anger, fear, ...) representative for music perception?
- should aesthetic emotions be distinguished from other emotions (guilt, shame, disgust, ...)?
- aroused vs. transported/evoked vs. conveyed moods?



#### ground truth data

- verbalization of emotions/moods usually misleading
- not easily *quantifiable*/categorizable
- changing over time?

### research questions

- are basic emotions (happiness, anger, fear, ...) representative for music perception?
- should aesthetic emotions be distinguished from other emotions (guilt, shame, disgust, ...)?
- aroused vs. transported | evoked vs. conveyed moods?



#### **■** ground truth data

- verbalization of emotions/moods usually misleading
- not easily *quantifiable*/categorizable
- changing over time?

### research questions

- are basic emotions (happiness, anger, fear, ...) representative for music perception?
- should aesthetic emotions be distinguished from other emotions (guilt, shame, disgust, ...)?
- aroused vs. transported/evoked vs. conveyed moods?

### mood recognition



classification into label clusters<sup>1</sup>

Cluster 1	Cluster 2	Cluster 3	Cluster 4	Cluster 5
Rowdy	Amiable/Good Natured	Literate	Witty	Volatile
Rousing	Sweet	Wistful	Humorous	Fiery
Confident	Fun	Bittersweet	Whimsical	Visceral
Boisterous	Rollicking	Autumnal	Wry	Aggressive
<b>Passionate</b>	Cheerful	Brooding	Campy	Tense/Anxious
		Poignant	Quirky	Intense
			Silly	

■ mood model, circumplex model

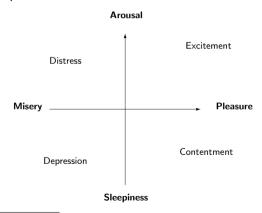
module 13: mood recognition 5

<sup>&</sup>lt;sup>1</sup>X. Hu and J. S. Downie, "Exploring Mood Metadata: Relationships with Genre, Artist and Usage Metadata," in *Proceedings of the International Society for Music Information Retrieval Conference (ISMIR)*, Vienna, 2007.

### mood recognition

Georgia Center for Music Tech Technology

- classification into label clusters
- mood model, circumplex model<sup>1</sup>



<sup>&</sup>lt;sup>1</sup> J. A. Russel, "A Circumplex Model of Affect," *Journal of Personality and Social Psychology*, vol. 39, no. 6, pp. 1161–1178, 1980, ISSN:

module 13: mood recognition 5

verview intro mood models **regression r**esults summary
○ ○ ○ ○ ○ ○ ○ ○ ○ ○

### mood recognition

mood model: regression modeling



### mapping

• (N-dimensional) observation (feature) to 2-dimensional coordinate (valence/arousal)

### training

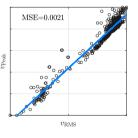
• find model to minimize error between data points and "prediction"

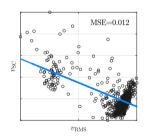
### regression regression

Georgia Center for Music Tech II Technology

■ linear regression: fit a linear function to a series of points  $(x_i, y_i)$ 

$$y_n = m \cdot x_n + b$$





■ other regression approaches: SVR, DNNs, etc.

# mood recognition range of results



- highly dependent on data
- **5 mood clusters**: 40–60% classification rate
- mood model: 0.1-0.4 absolute prediction error (unit circle)

# mood recognition range of results



- highly dependent on data
- 5 mood clusters:

40-60% classification rate

- mood model:
  - 0.1–0.4 absolute prediction error (unit circle)

### summary lecture content



#### emotion and mood

- emotion: temporary, related to external stimuli
- mood: long term, diffuse affective state

#### features

baseline features are identical to genre and similarity tasks

#### inference

often done as regression (as opposed to classification)

