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# The role of F0 trajectory in the emotion identification

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# Two emotional distinction theories

- **The discrete emotion theory**

- Basic discrete emotions exist:

(1) surprise, (2) interest, (3) joy, (4) rage (5) fear (6) disgust (7) shame (8) anguish

- Individual emotions have biological and neurological profiles

- **The dimensional theory**

- Two emotional dimensional spaces distinguish emotions

(1) **valence** – how positive or negative an emotion is

(2) **arousal** – the intensity of an emotion

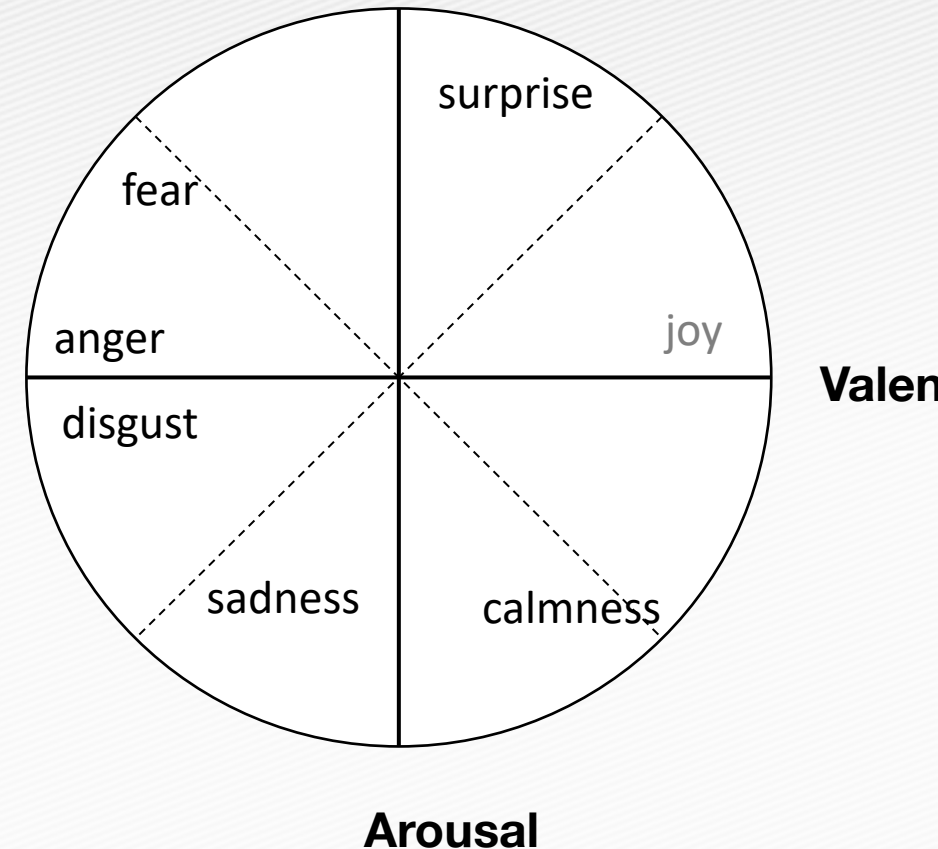
# The discrete emotion approach

- Emotions are discrete, measurable, and physiologically distinct.
- Certain emotions appeared to be universally recognized.

→ Many studies have examined the vocal characteristics of speech in hope of defining a vocal signature for each basic emotion (Russell 2003)

# The Dimensional approach

- The strongest single association found for vocal acoustic have been with the sender's general **arousal** level.
- High-arousal emotions such as **anger** and **joy** have similar characteristics low arousal emotions such as **sadness**
  - greater loudness,
  - higher pitch
  - faster speech
- Few works have concentrated on distinguishing emotions between positive- and negative- valence emotions such as **anger** and **joy**.



# Research topic

- F0 contours contains discriminatory information about emotions.
- Very few can be found in the literature that made the efforts to describe the shape of f0 contours directly in classifying emotions

# The Ryerson Audio-Visual Database of Emotional Speech and Song (RAVDESS)

- The RAVDESS dataset is a multimodal validated English dataset that contains speech, song, and video files that represent 8 emotions.
- The portion of the dataset that I use in this study is the speech audio files that are represented by 1440 wave file.
- Twenty-four professional actors (12 female and 12 male) with 60 trials for each actor produced the 1440 wave files ( $24 \times 60 = 1440$ ).

# The Ryerson Audio-Visual Database of Emotional Speech and Song (RAVDESS)



- The actors vocalized two sentences in a neutral North American accent.
  - “Kids are talking by the door”
  - “Dogs are sitting by the door”
- The emotions
  - neutral, calm, happy, sad, angry, fearful, surprise, and disgust
- Each expression is produced at two levels of emotional intensity (normal and strong) except for the neutral emotion that is recorded in a normal intensity only.

# Generalized Additive Mixed Modeling

- In Linear Model, the mean of data is modeled as a sum of linear terms

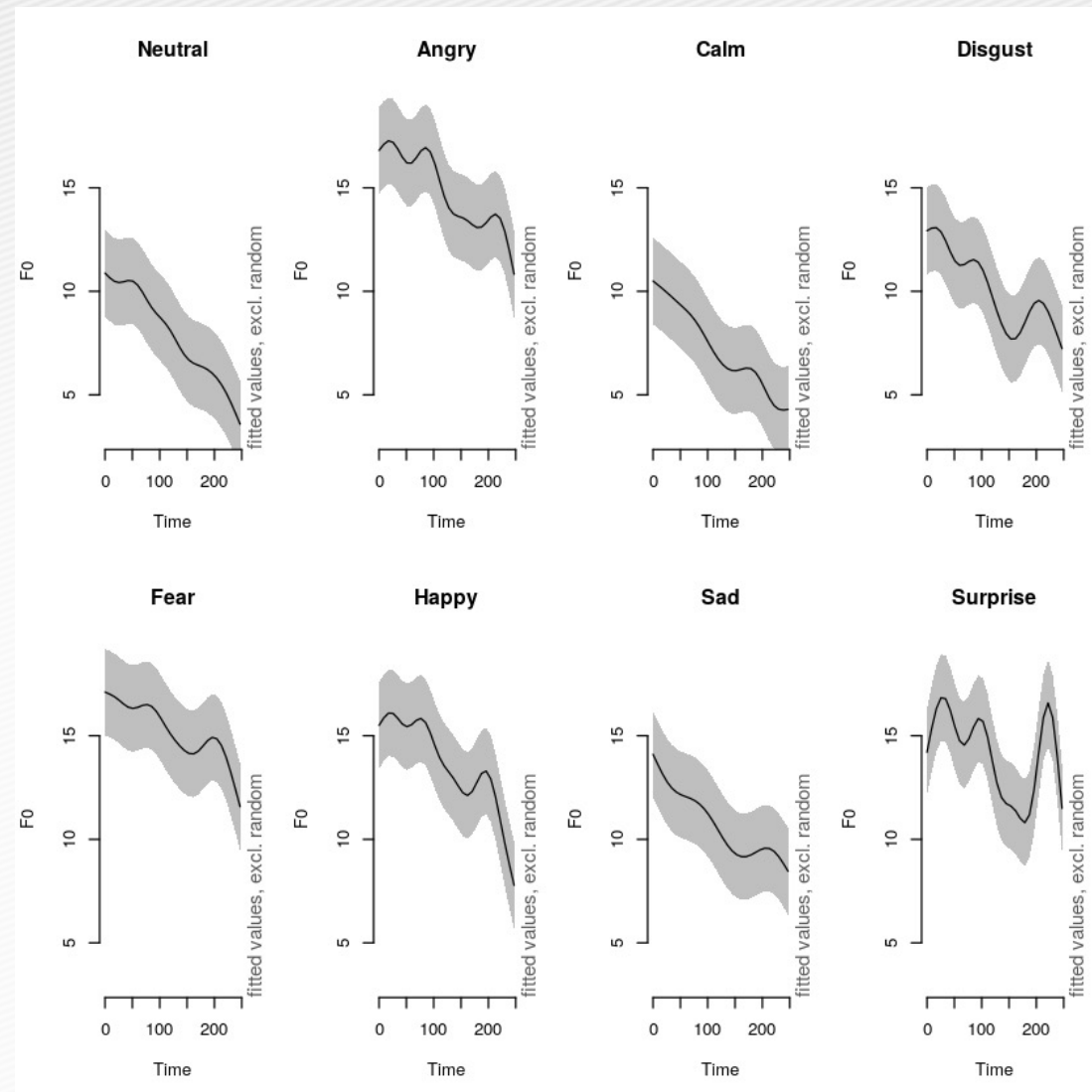
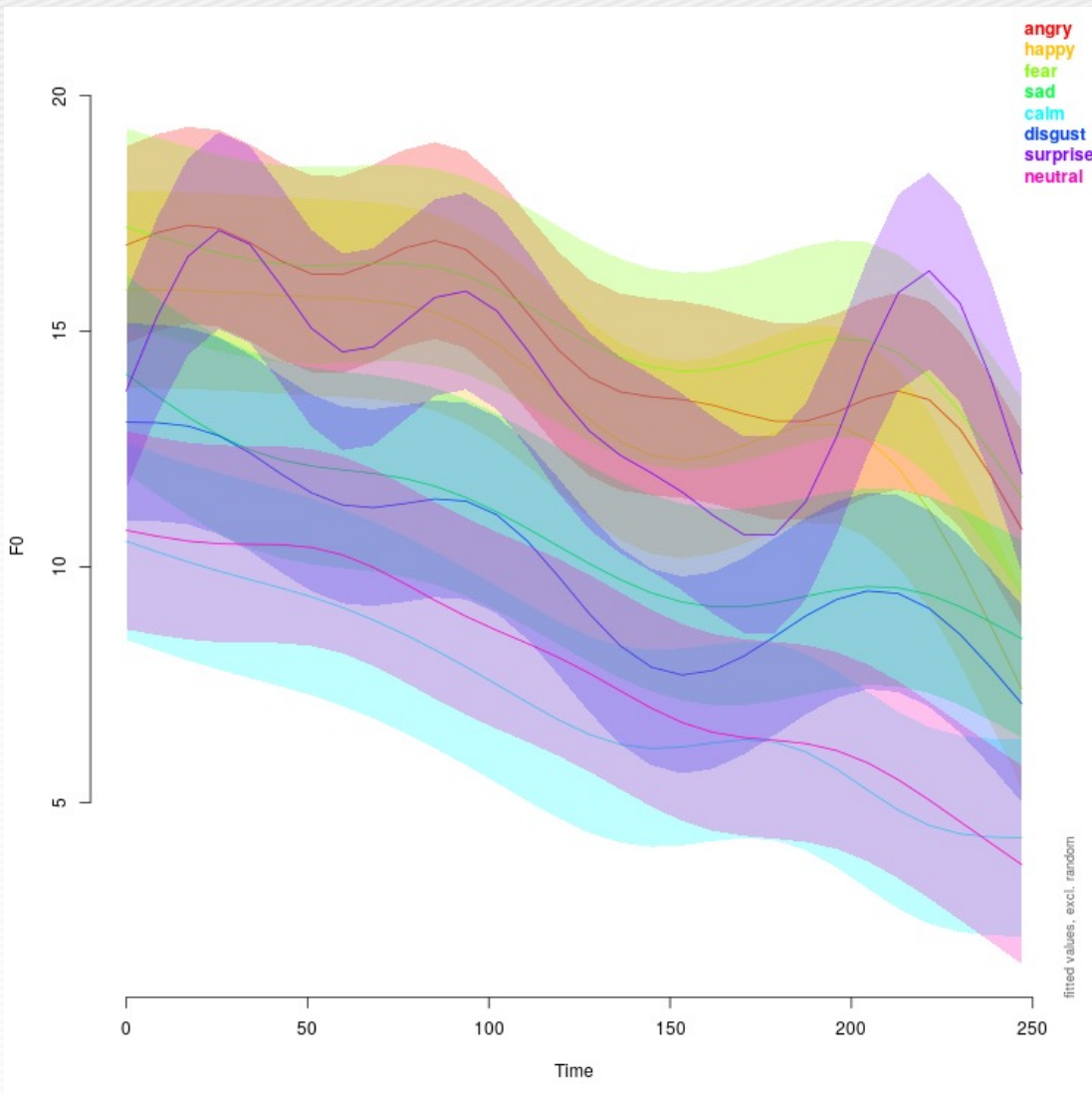
$$y_i = \beta_0 + \sum_j \beta x_{ji} + \varepsilon_i$$

- In Generalized Additive Mixed Model, the mean of data is modeled as a sum of **smooth** functions (= smooths)

$$y_i = \beta_0 + \sum_j s_j(x_{ji}) + \varepsilon_i$$



# GAMM approach to the F0 contour modeling



# Gamm Modeling

Formula:

```
F0 ~ Emotions + s(Time, by = Emotions, k = 10) + s(Actor, bs =
"re") +
  s(Actor, Emotions, bs = "re")
```

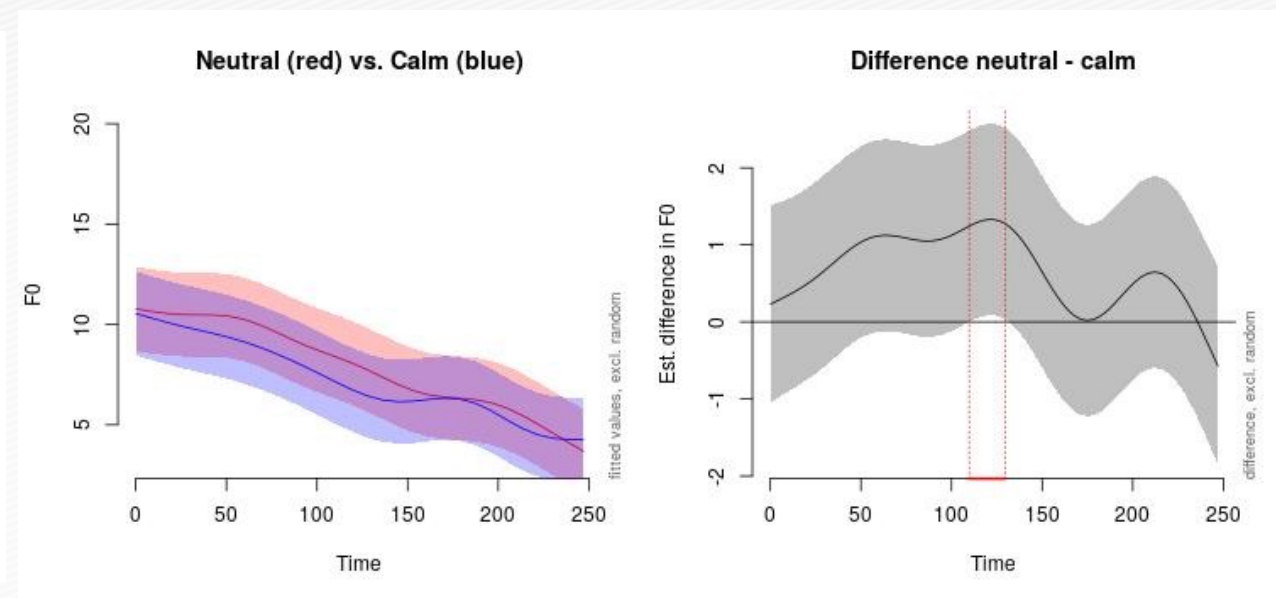
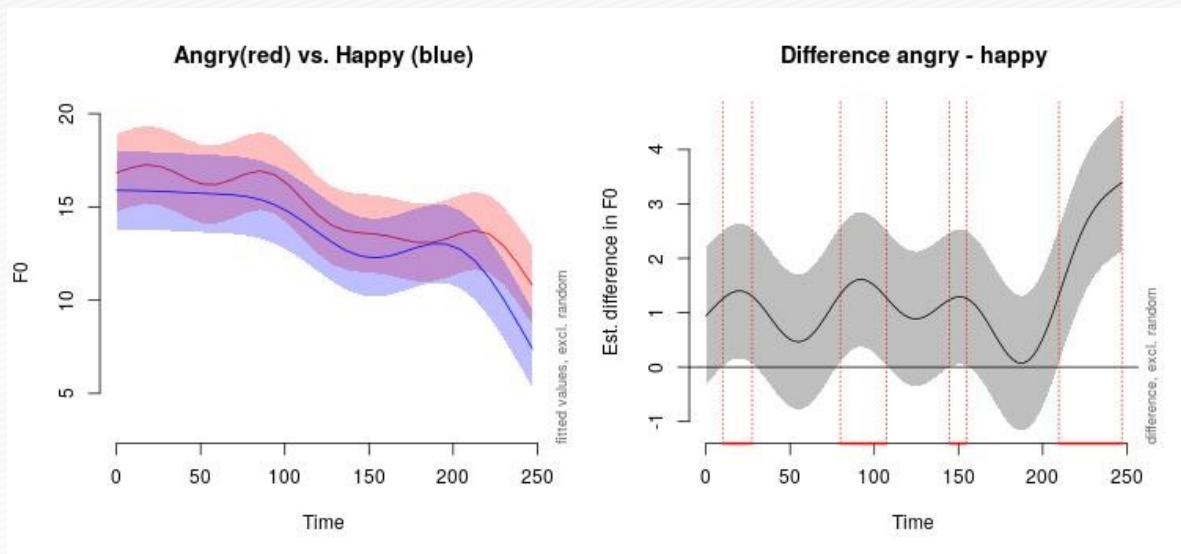
Parametric coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	7.9017	1.0637	7.429	1.10e-13	***
Emotionsangry	6.9618	0.6277	11.091	< 2e-16	***
Emotionscalm	-0.6860	0.6277	-1.093	0.274459	
Emotionsdisgust	2.2108	0.6277	3.522	0.000428	***
Emotionsfear	7.3391	0.6277	11.692	< 2e-16	***
Emotionshappy	5.7515	0.6277	9.163	< 2e-16	***
Emotionssad	2.8123	0.6277	4.480	7.46e-06	***
Emotionssurprise	6.2753	0.6277	9.997	< 2e-16	***

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 Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

R-sq.(adj) = 0.649    Deviance explained = 64.9%  
 fREML = 1.0445e+06    Scale est. = 20.228    n = 357120

# Pair-wise comparison of contours



# Conclusion

- I attempted to model Emotions using F0 contours as an input to generalized additive model (GAM)
- The present approach has predictive power (64.9%).
- The additive model provides visualized aids and makes us better understand validity data obtained from human labelers.

# THANK YOU

