

Music Wellness Center, Ewha University, South Korea | June 26, 2024

Neural encoding of musical emotion

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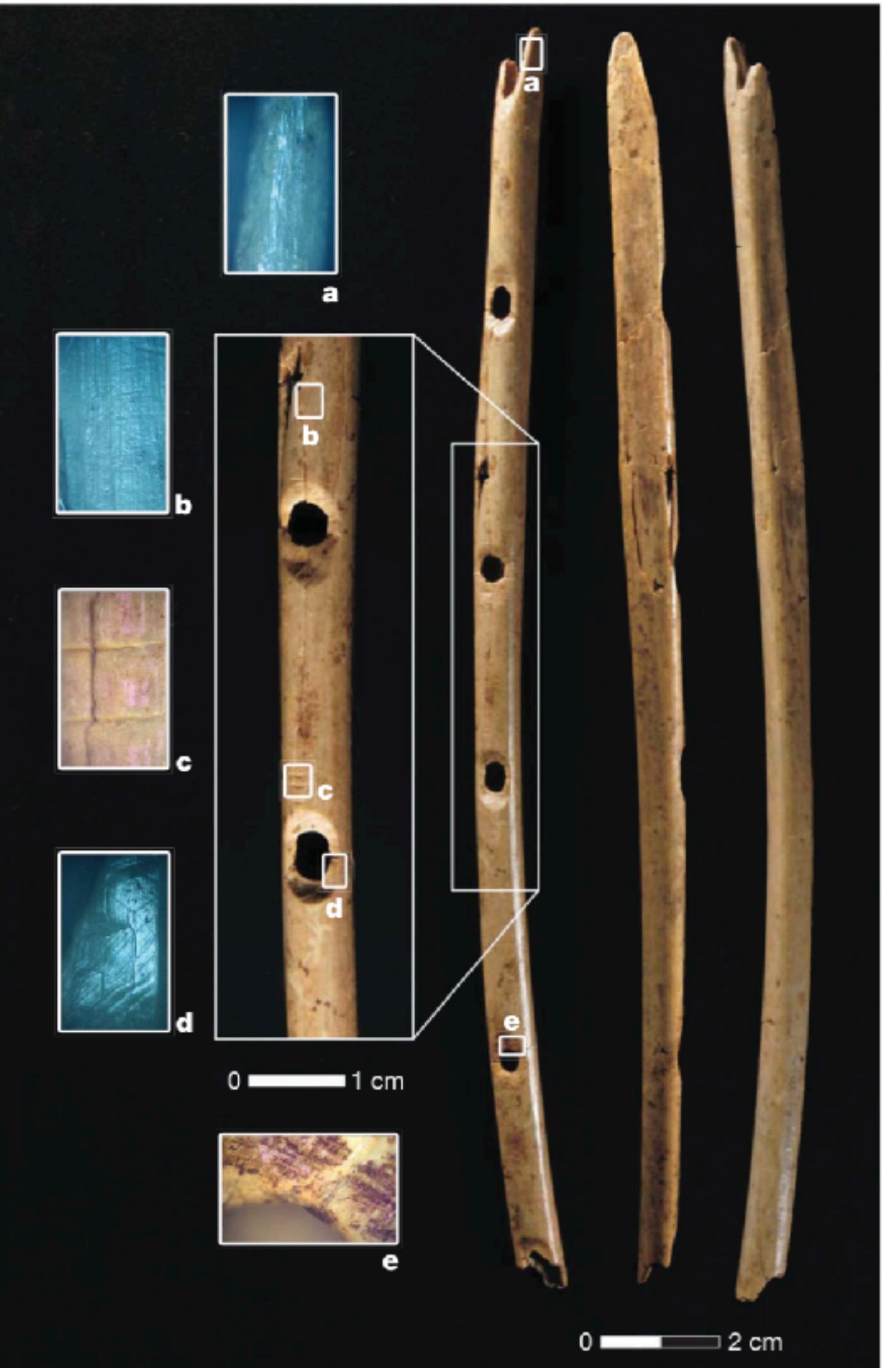
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²Department of Psychology & Neuroscience, Duke University, NC, USA



Music in culture

- Current archeological evidence:
 - Ancient origin of music: > 35k yr ago [1]
 - Human & speech: 200k yr ago [2,3]
 - Writing: 5k yr ago [4]
 - Ubiquitous in every human culture [5]

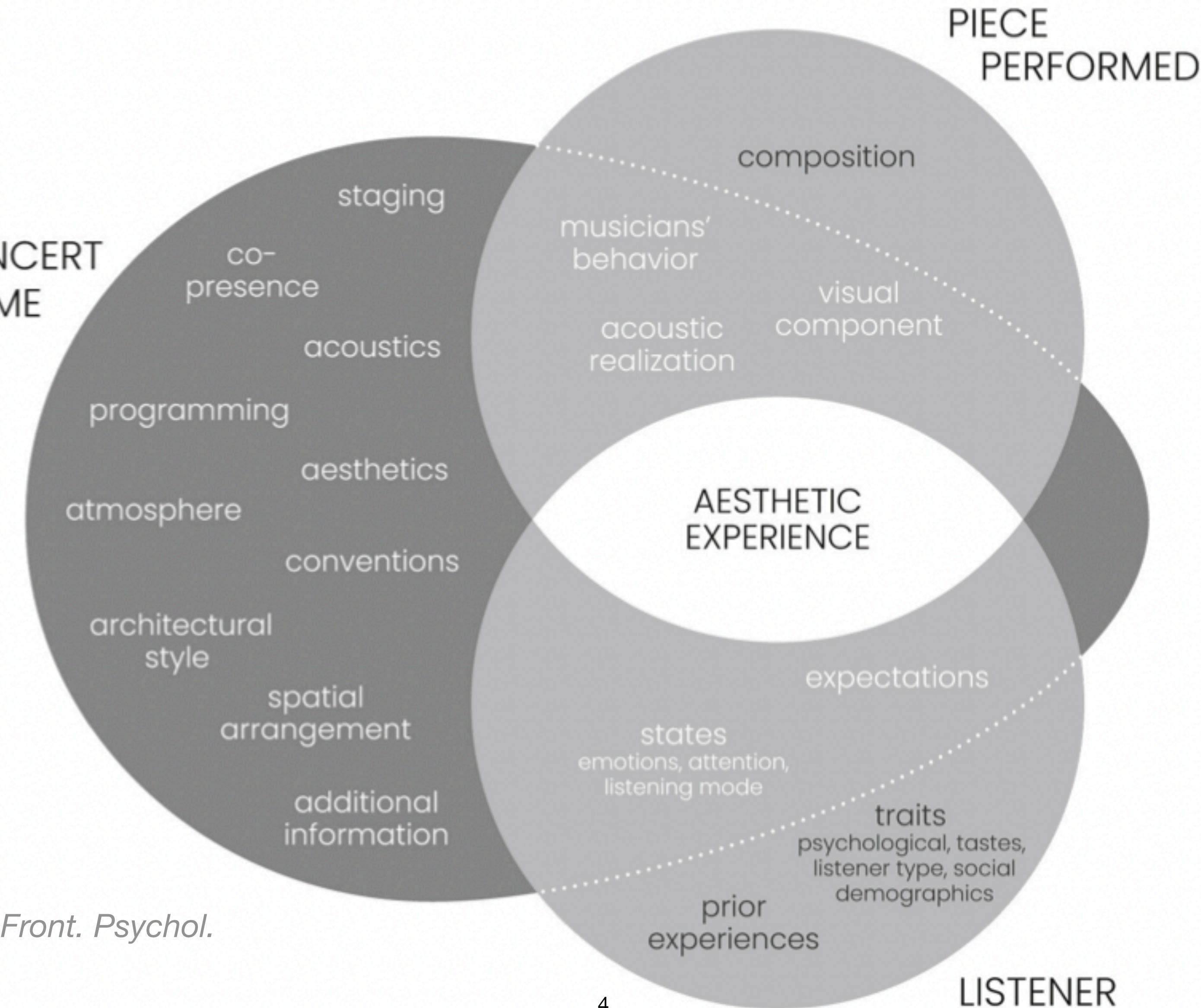


[1] Conard et al. (2009). Nature. [2] Mounier & Lahr. (2019). Nat Comm.

[3] Perreault & Mathew. (2012). PLOS ONE. [4] Senner. (1991). [5] Honing et al. (2015). Phil Trans B.

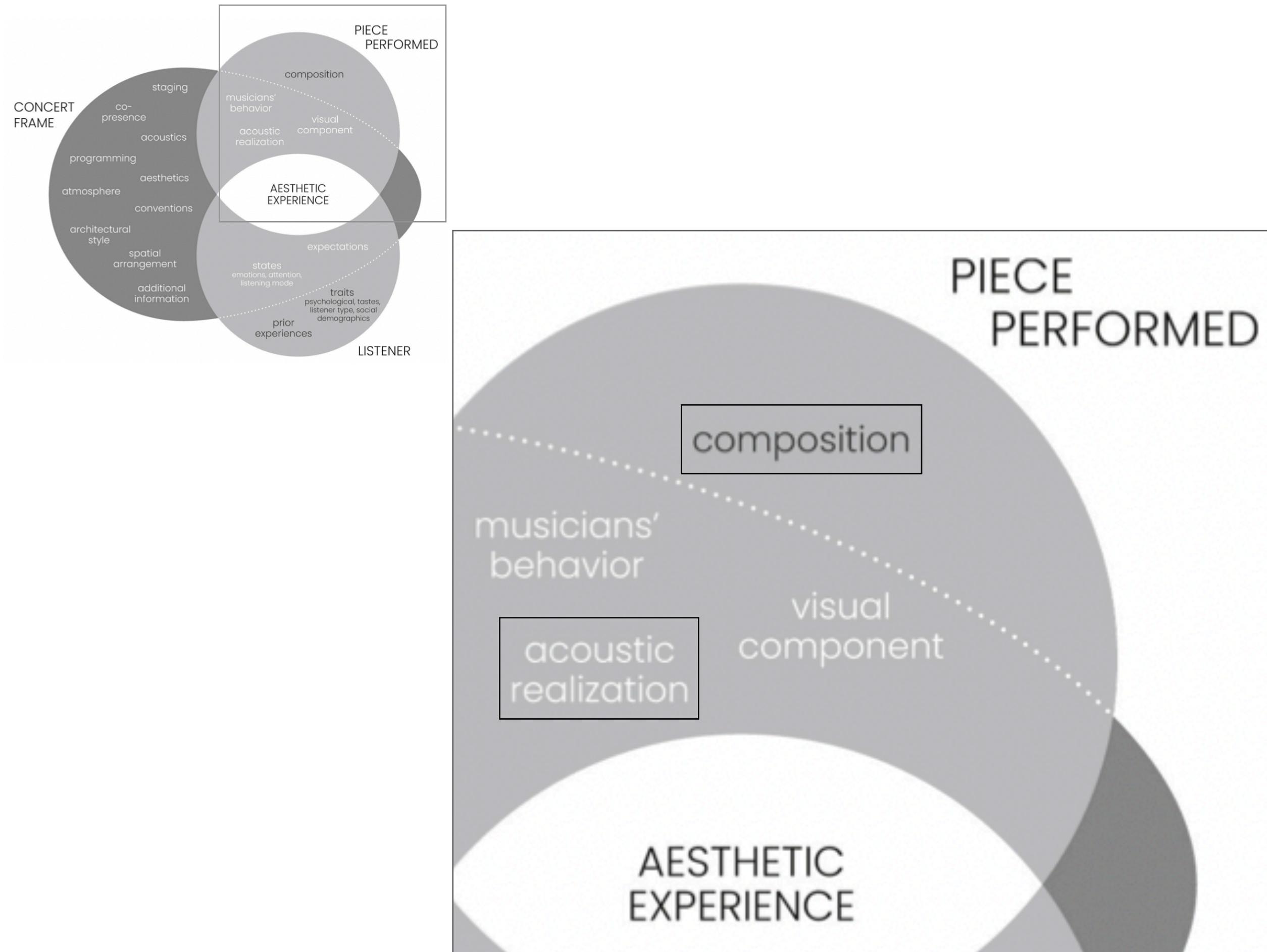
35,000 year-old vulture bone flute
from Danube, Germany. [1]

Frameworks of music-evoked emotion

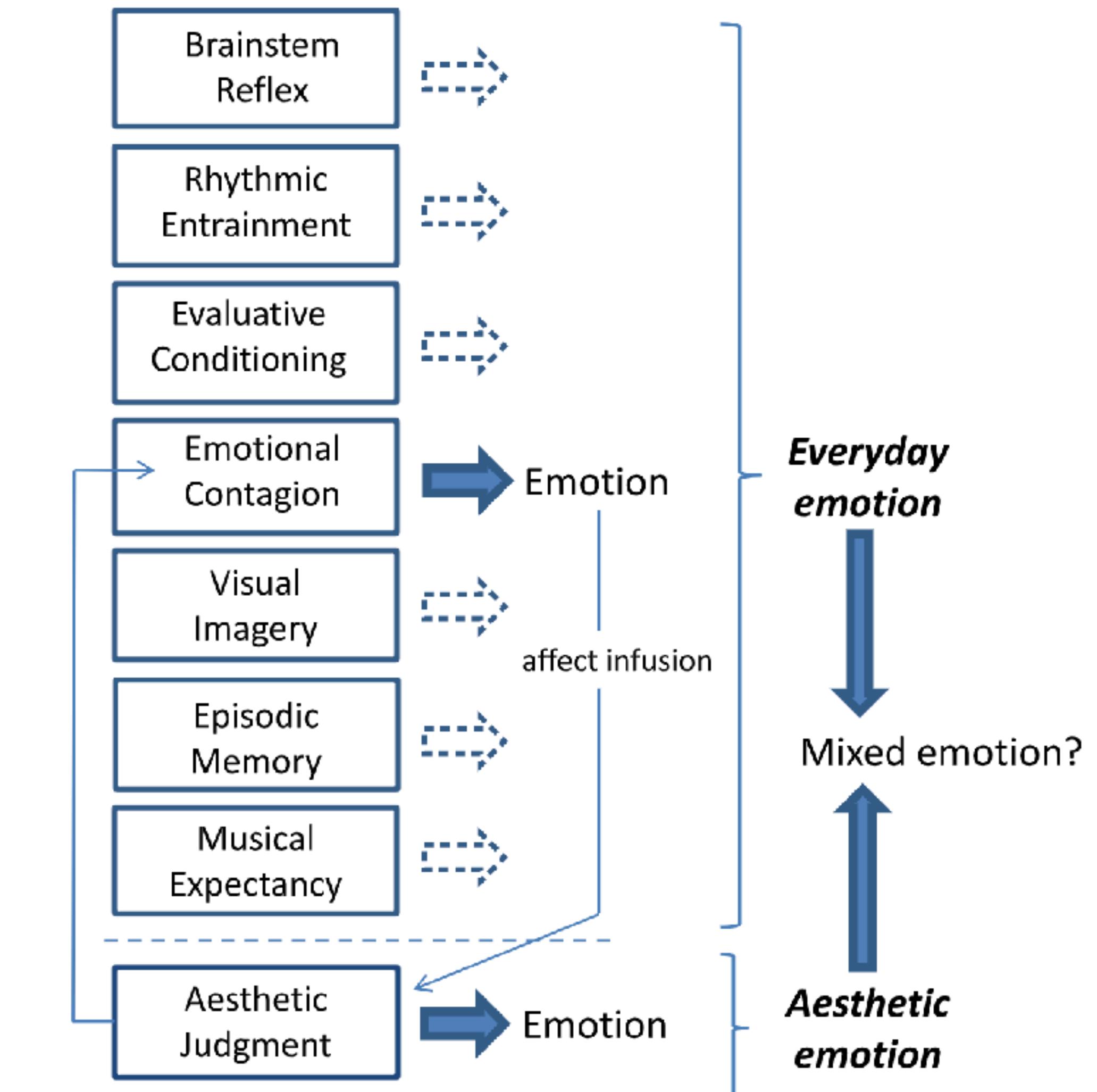


Wald-Fuhrmann et al., 2021, *Front. Psychol.*

Frameworks of music-evoked emotion



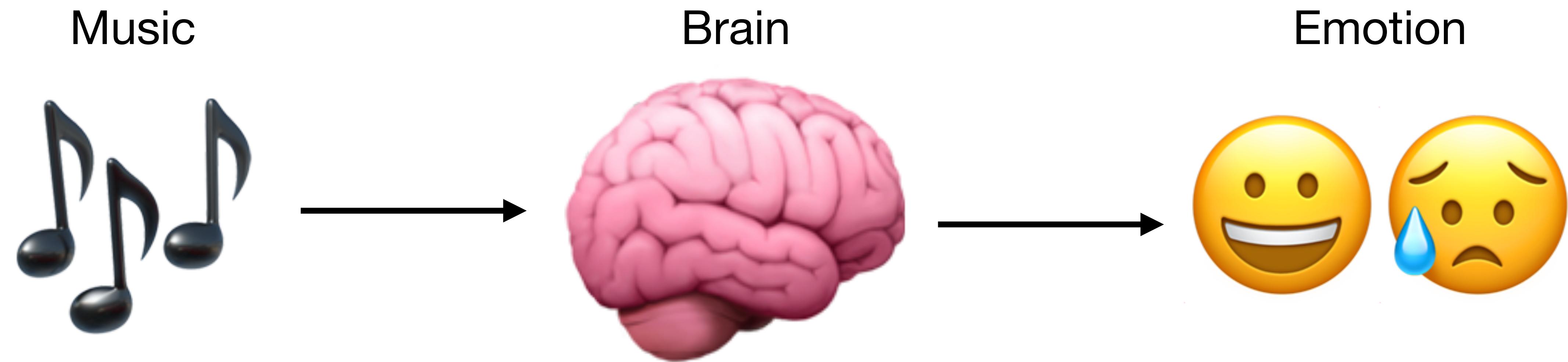
Wald-Fuhrmann et al., 2021, *Front. Psychol.*



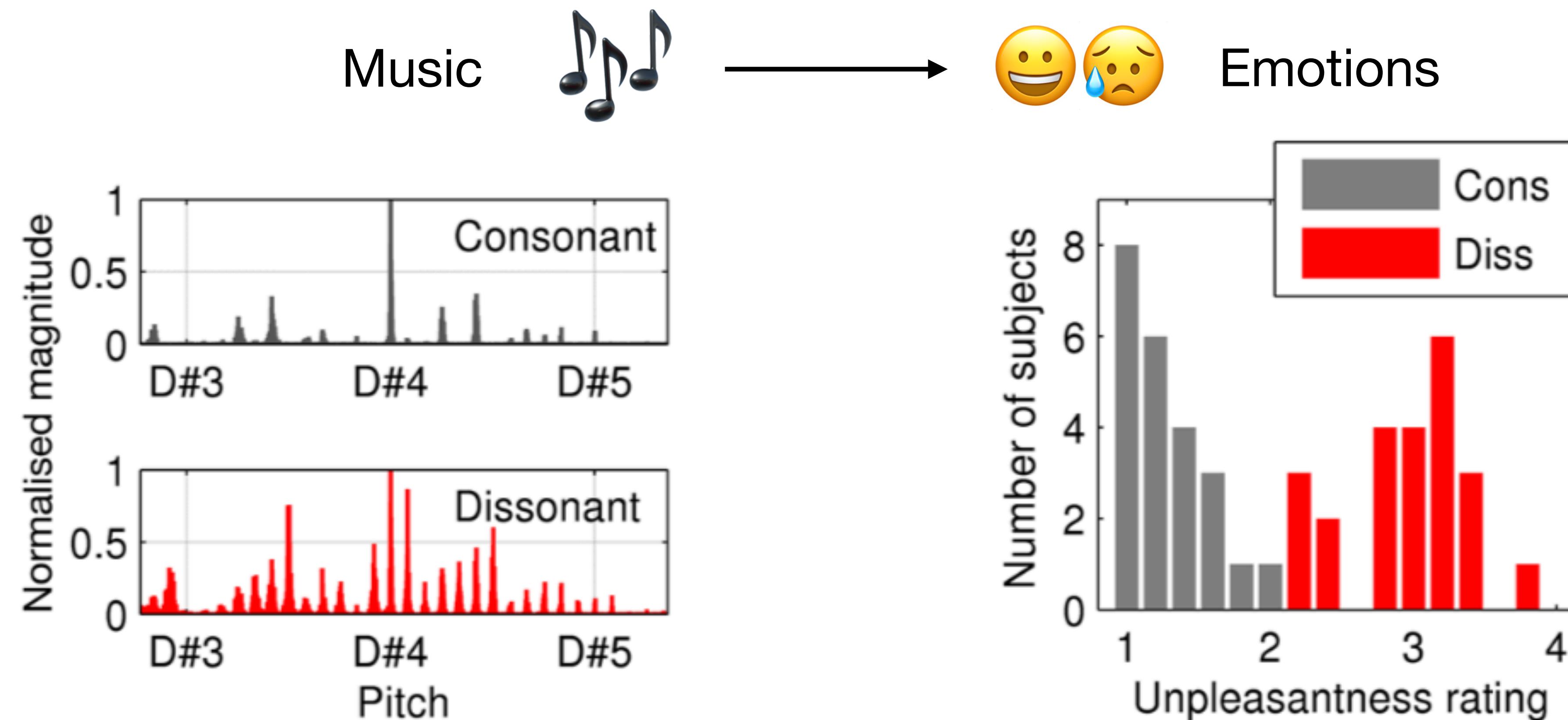
"BRECVEMA" (Juslin, 2013, *Physics of Life Rev.*)

How does music evoke emotions *via* the brain?

Neuroscientific view



Study I: Consonant vs. Dissonant music



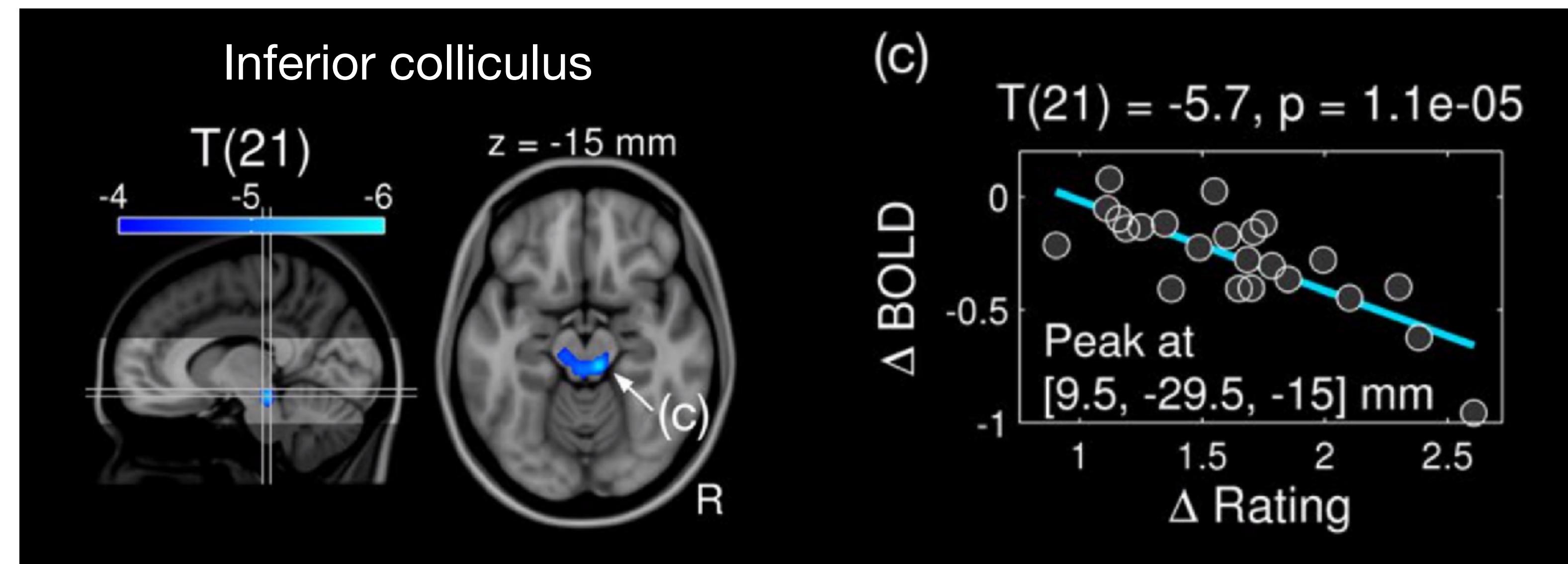
“Dissonant” stimuli were rated more unpleasant.

stimuli per condition = 20, stimulus duration = 30 s, # subjects = 23 (25.9 yo; 13 F)

Study I: Consonant vs. Dissonant music



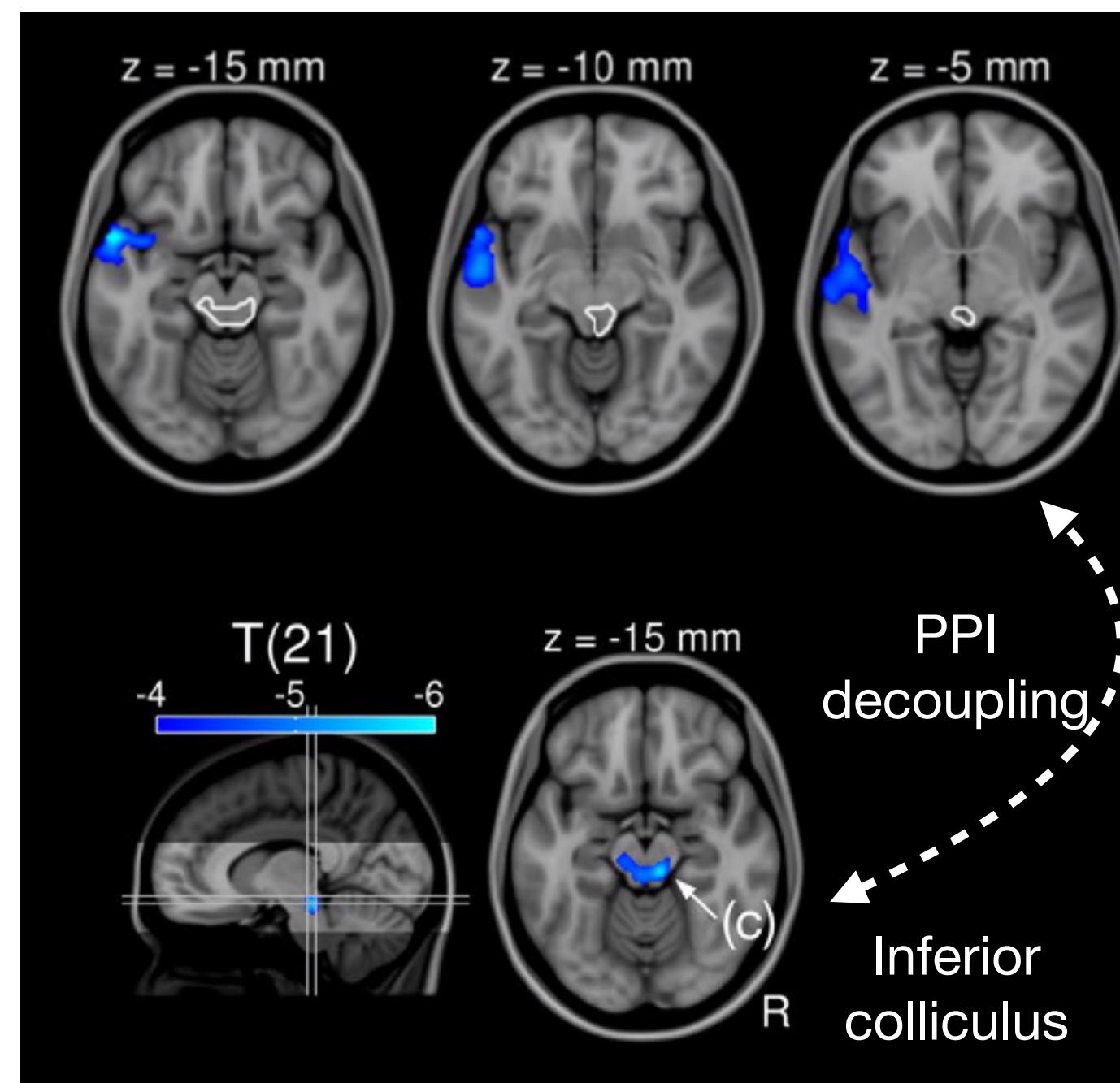
$$\Delta BOLD = \beta_0 + \beta_1 \Delta Rating + error$$



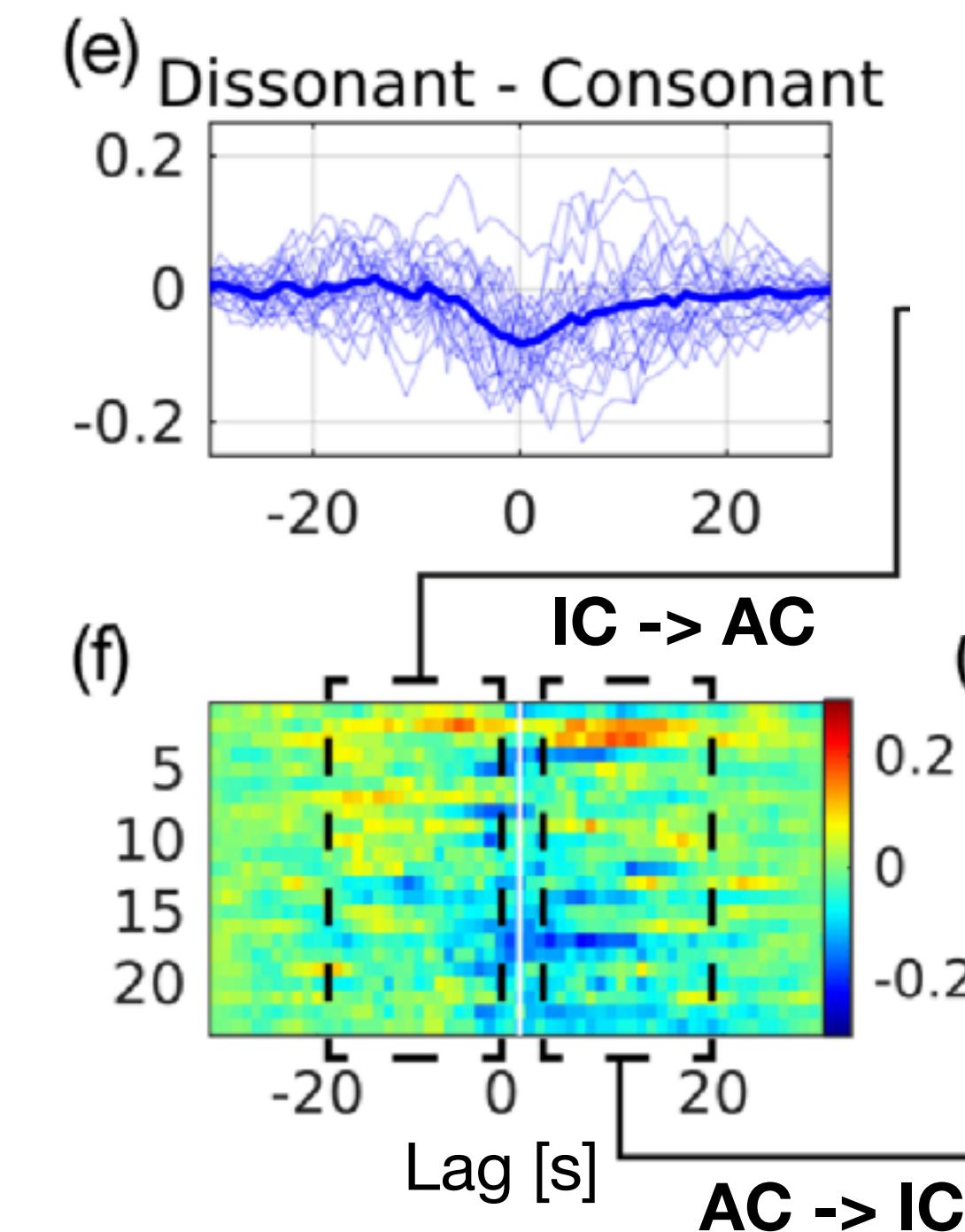
BOLD activation in the inferior colliculus (IC) decreased more in individuals who rated dissonant versions worse.

Study I: Consonant vs. Dissonant music

Psycho-Physiological Interaction (PPI)



Cross-correlation



The IC decoupled more from the left auditory cortex (top-down) in individuals who rated dissonant versions worse.

Limitations of controlled stimuli

Ceteris paribus... and the subject is not listening.

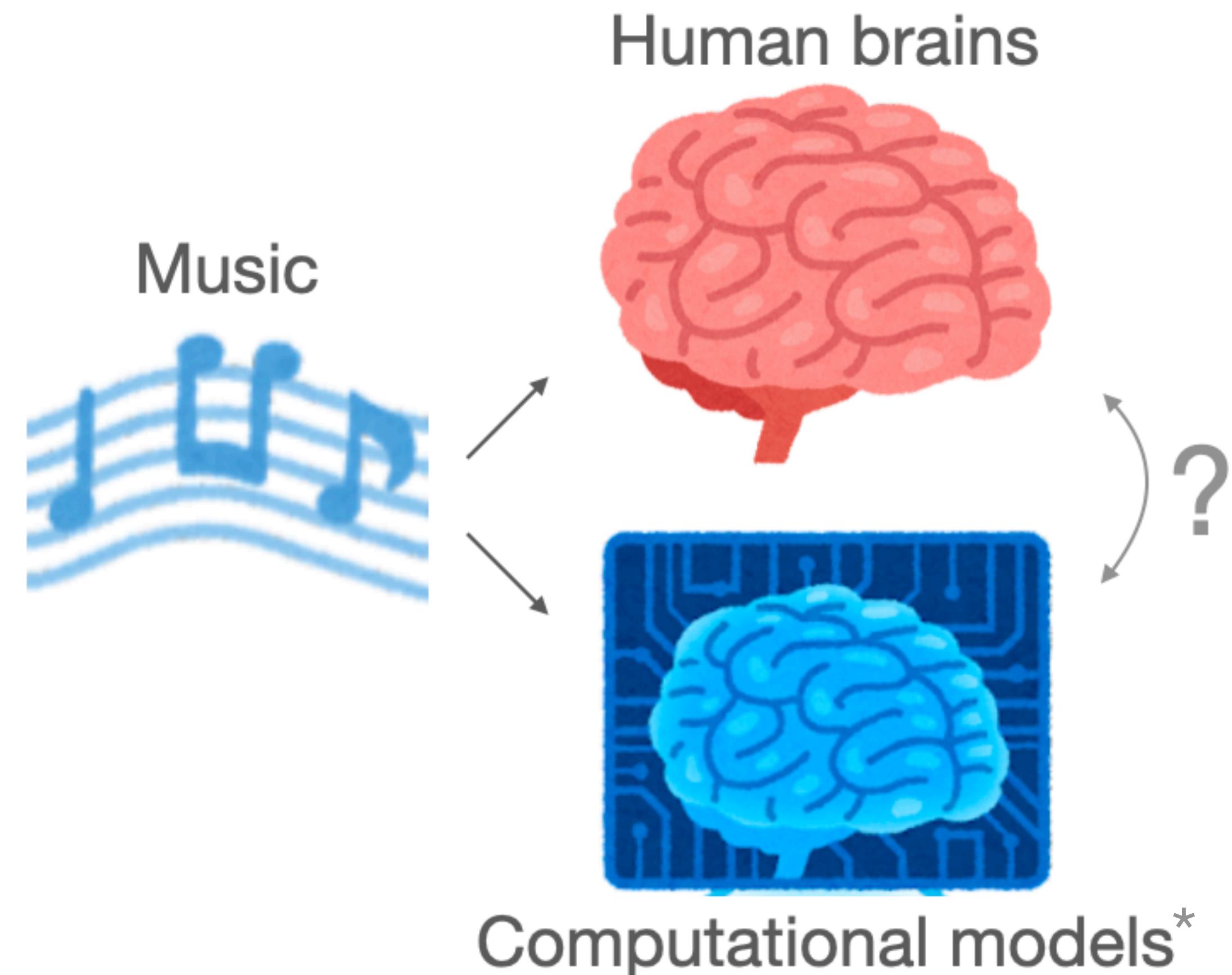
Disruptive manipulations: validity, generalizability, & specificity?



Solution: Use of natural(istic) stimuli to increase the external validity of neuroscientific studies.

Study II: Approach

Computational models to extract information from natural music



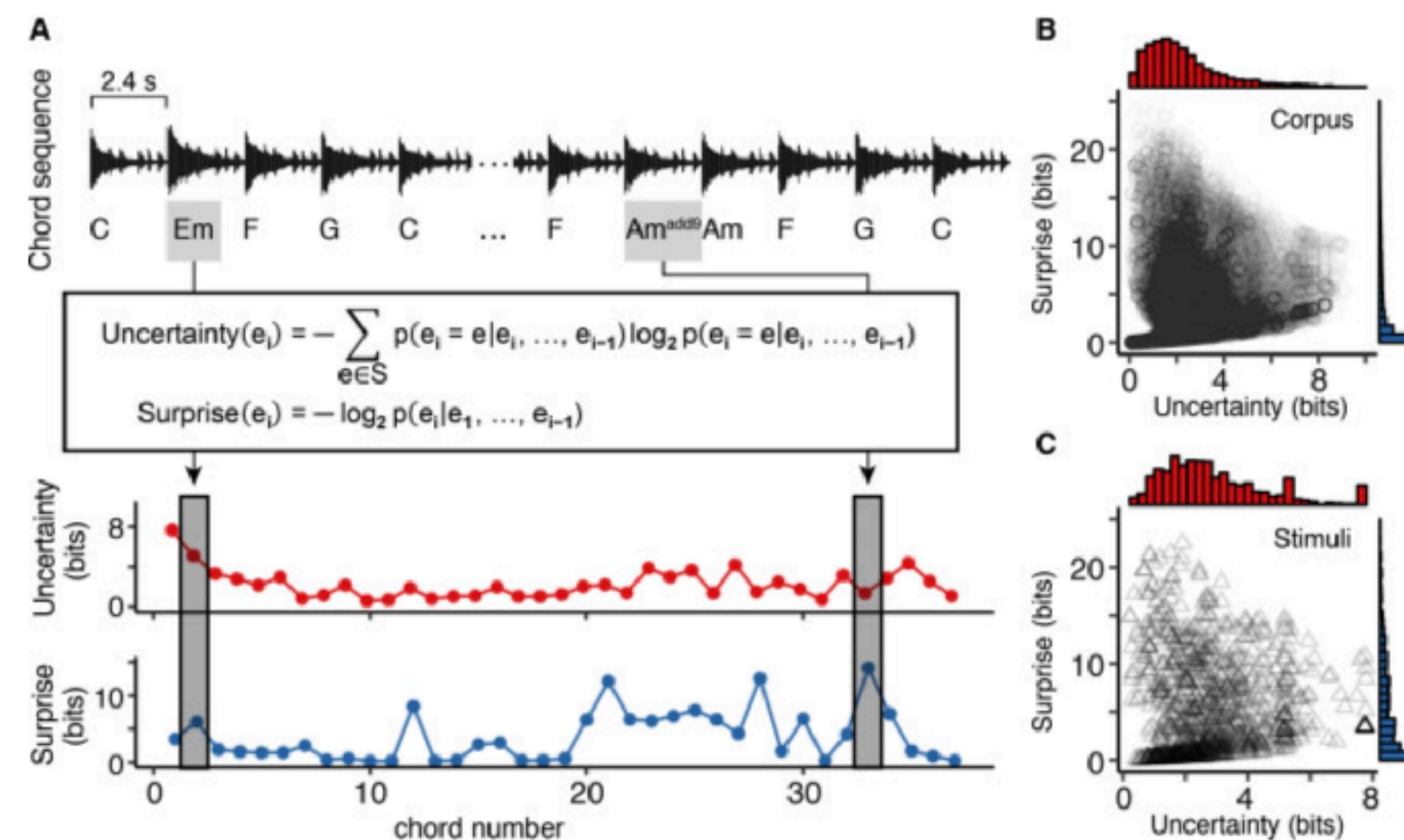
*Possible mid/high-level representations that are relevant to emotions

How can we extract "meaningful" features from music signals?

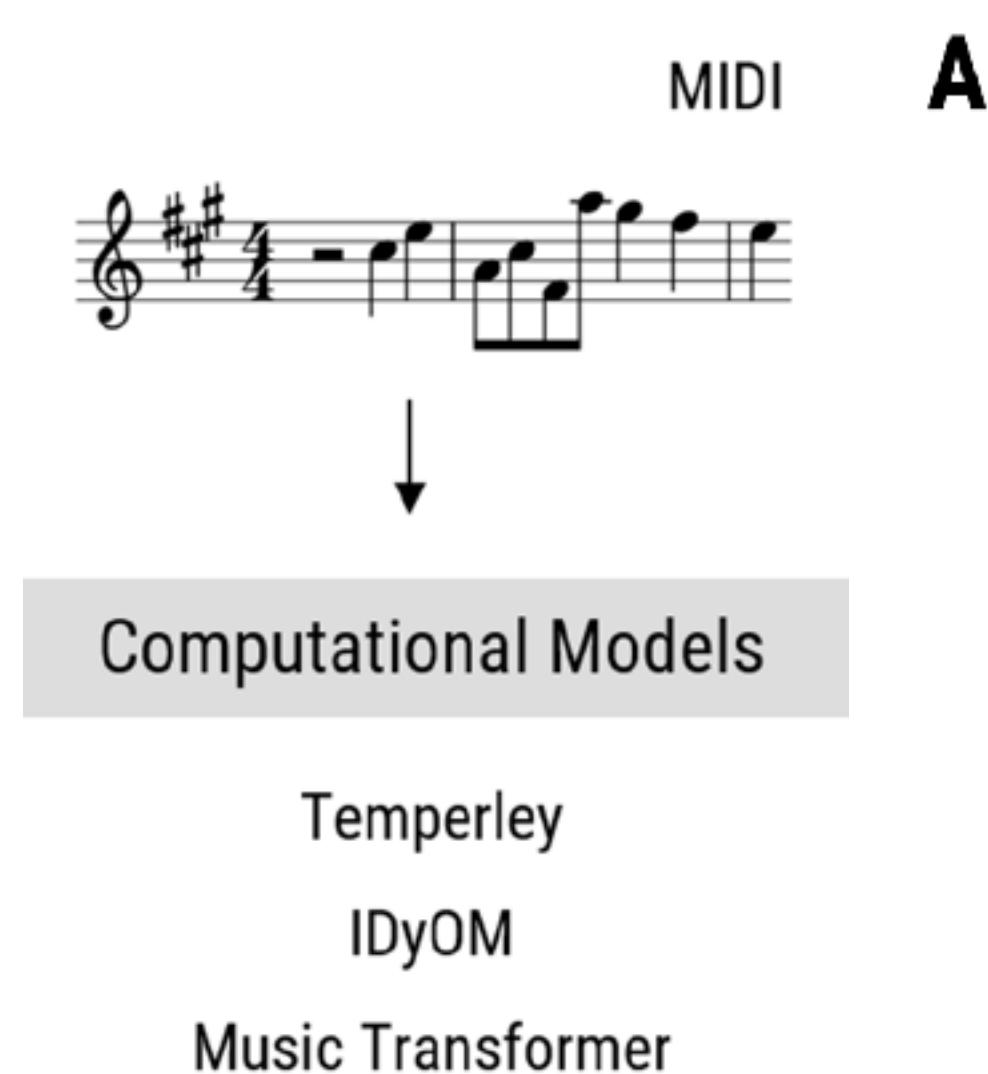


Symbol-domain models

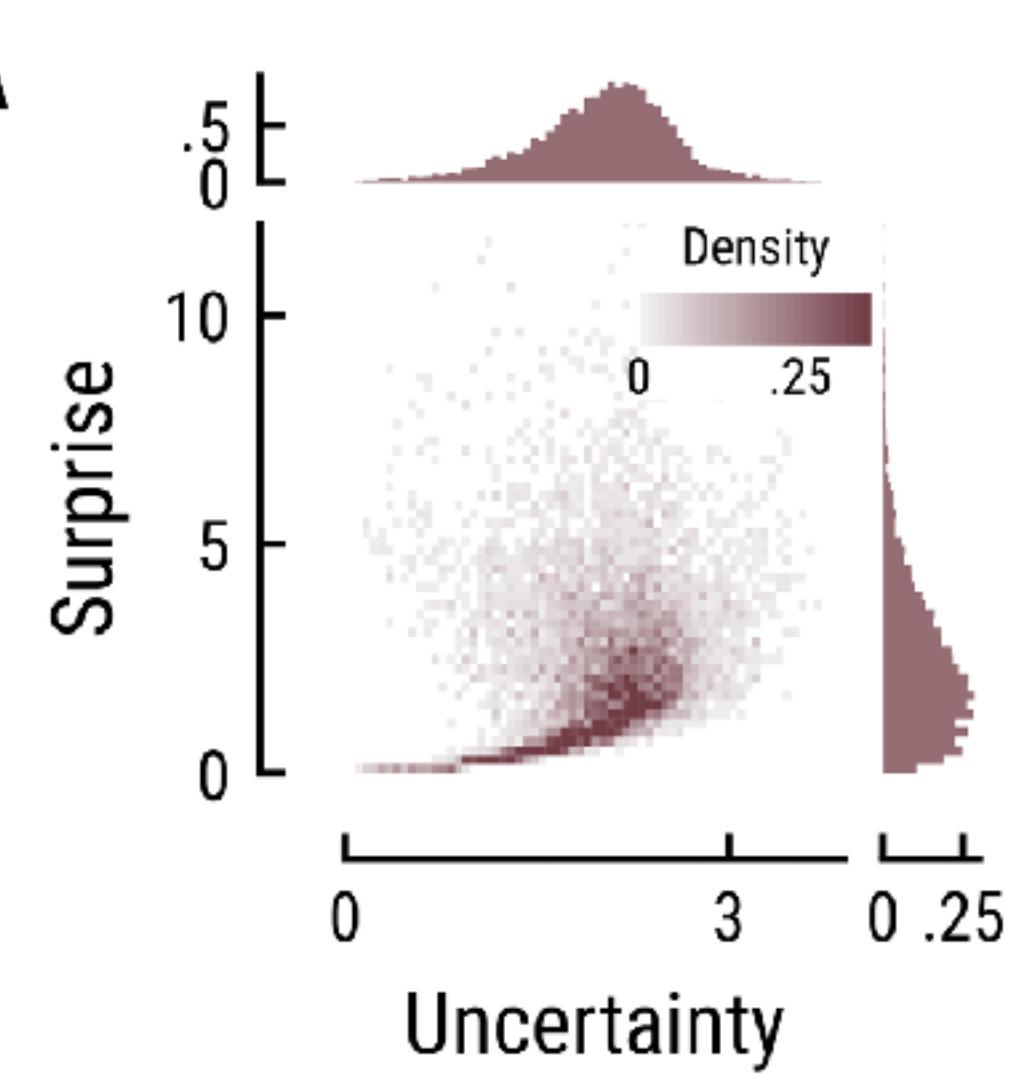
Modeling on MIDI notes: **n-gram modeling** (e.g., PPM, HMM),
deep neural networks (e.g., Melody RNN, MusicRNN, Music Transformer)



Cheung et al., 2019, *Current Biology*.



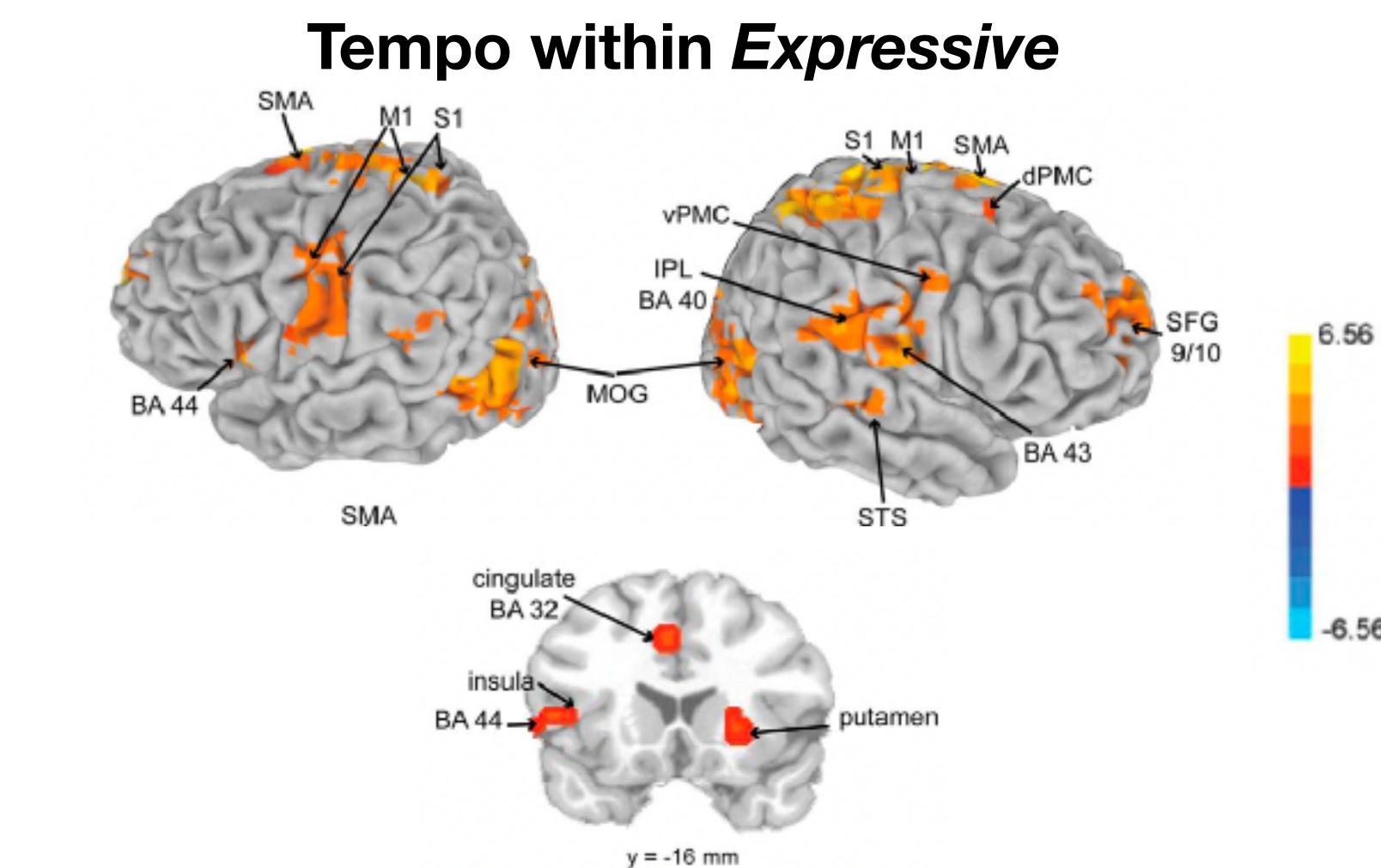
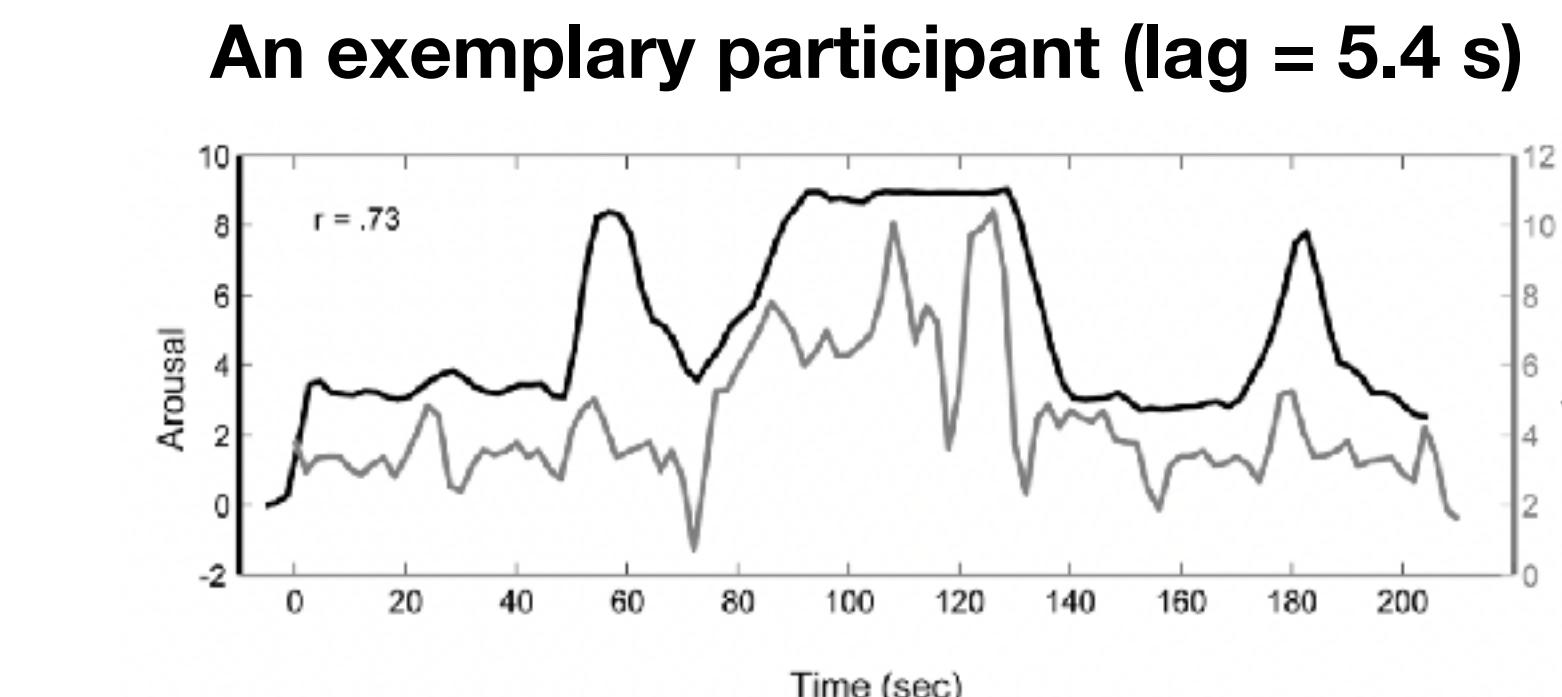
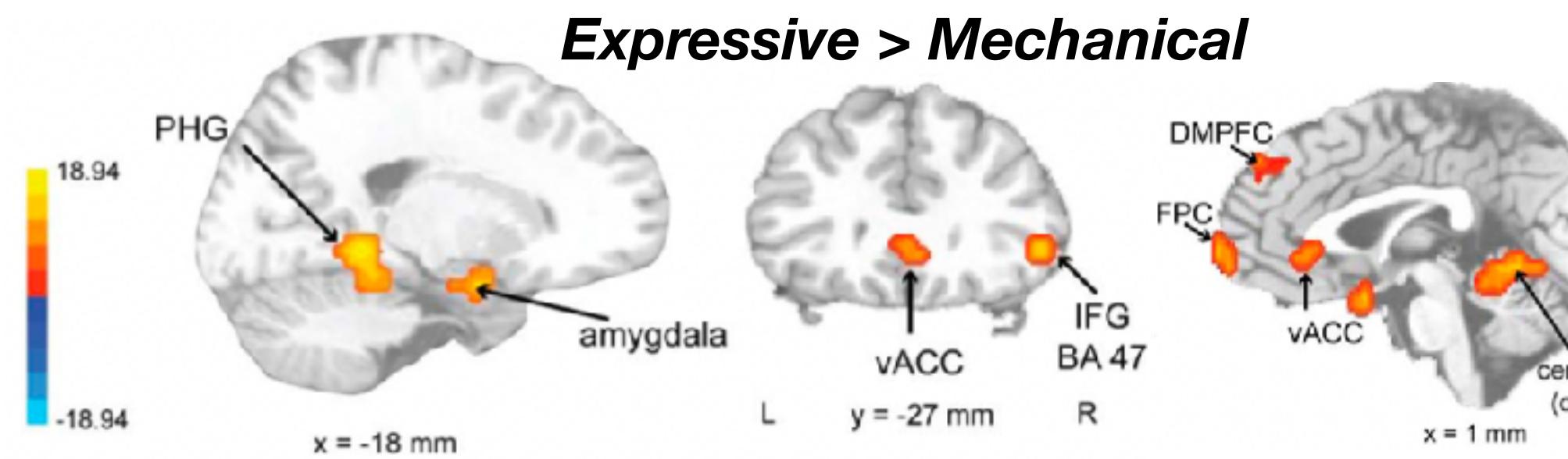
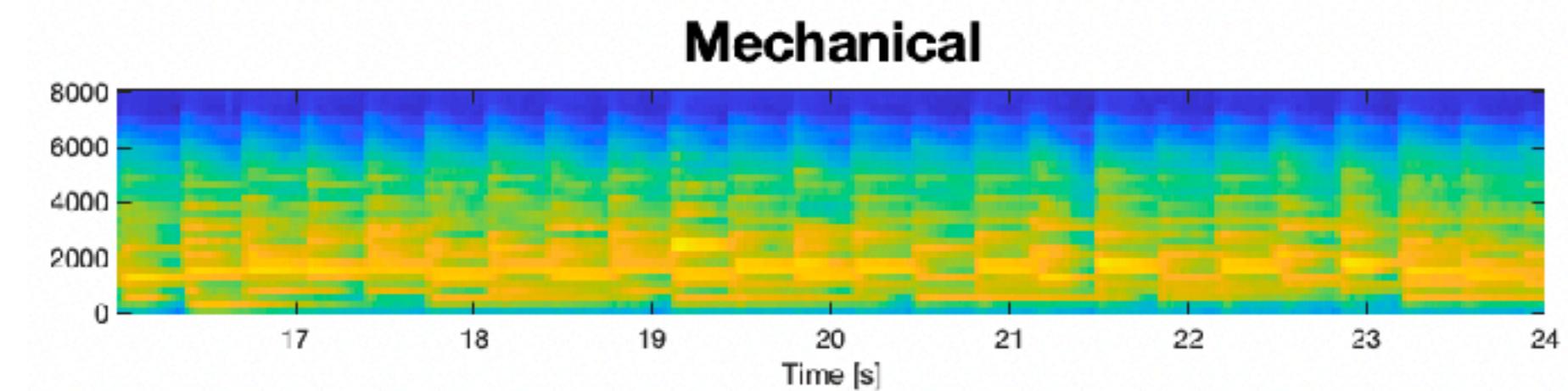
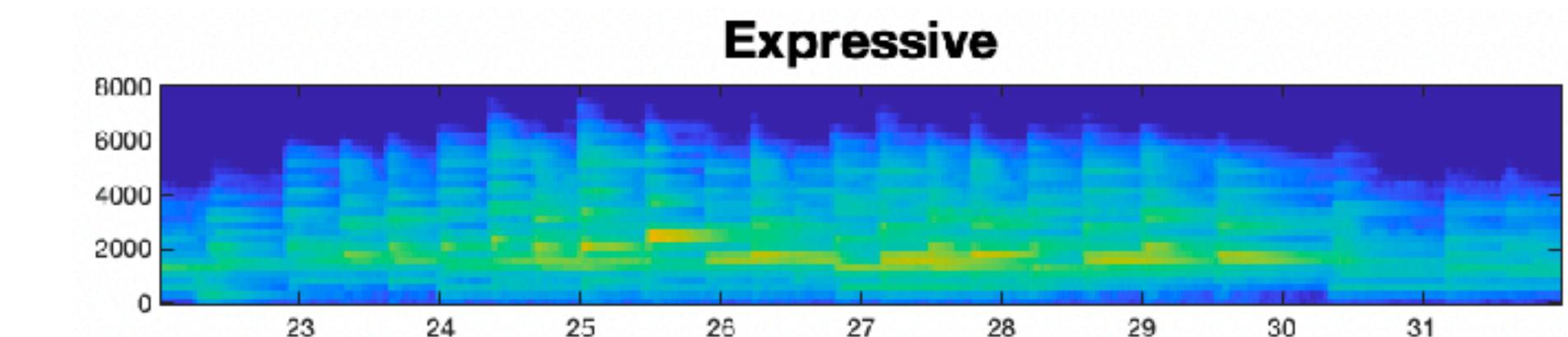
Kern et al., 2022, *eLife*.



But notes/chords should be transcribed correctly.

Effects of expressions (tempi & dynamics)

Chapin et al., 2010, *PLOS One*.

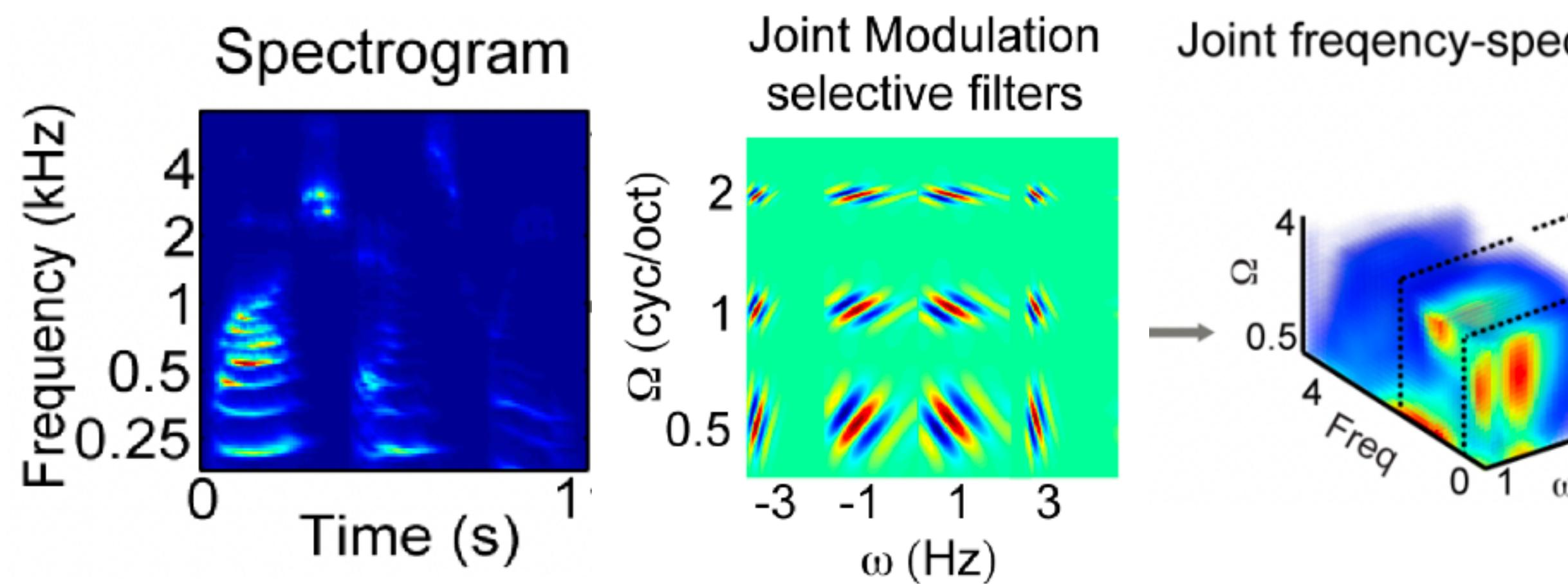


Chapin et al., 2010, *PLOS One*.

Symbolic models do not capture musical expressions!

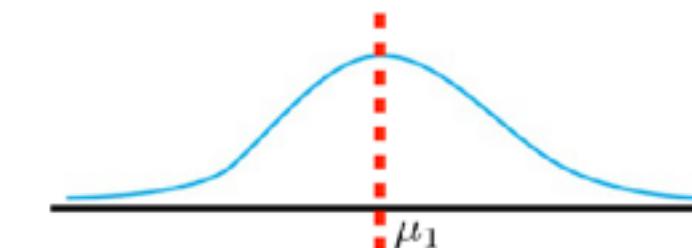
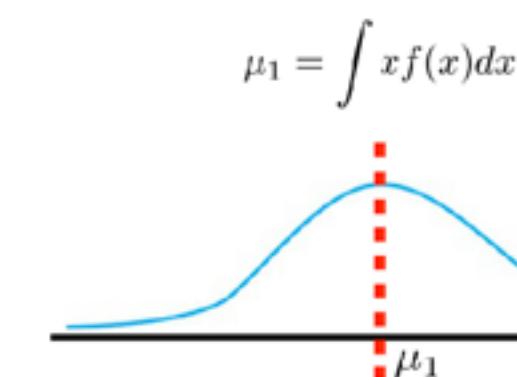
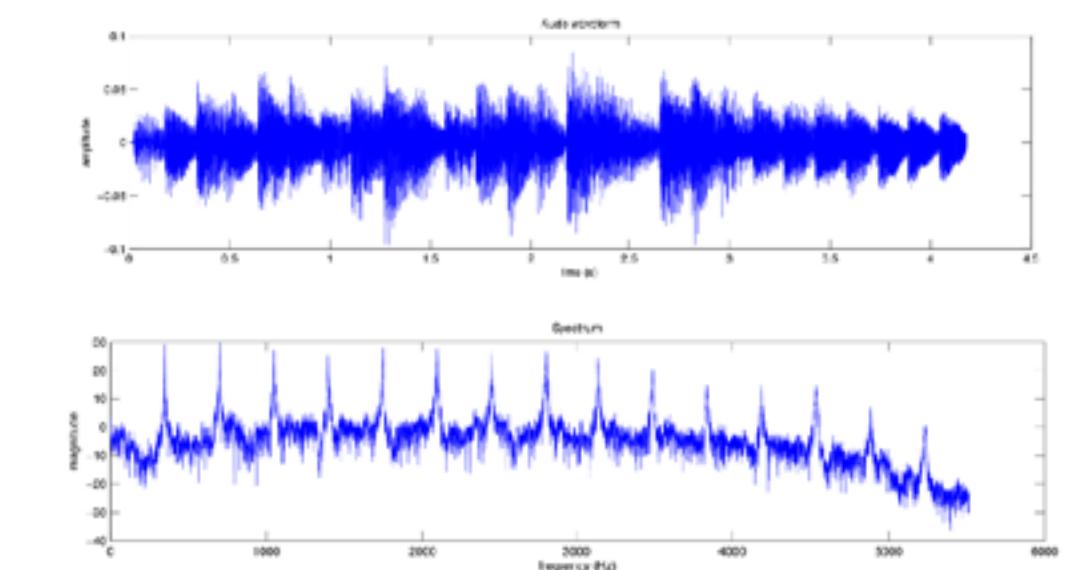
Audio-domain models

Auditory models: simulated activity of peripheral and central auditory neurons



Santoro et al., 2014, *PLOS Comp Biol.*

Music information retrieval (MIR) models: acoustic features relevant to music database



Lartillot, 2017, *MIRtoolbox User's Manual*.

Traditional models mostly simulate cochlear and primary auditory cortex activity.



The image shows a YouTube video player interface. The main video frame features a man with dark hair and glasses, wearing a light-colored shirt, speaking. In the foreground, a large white rectangular box contains the text "Mushin lurning" in a bold, sans-serif font. Below the video frame, the YouTube control bar includes a play button, volume icon, and timestamp "1:22 / 4:03". To the right of the video frame are standard YouTube controls for volume, captions, settings, and sharing. At the bottom of the player, there is a title card reading "Interview with a Postdoc, Junior Python Developer". Below the title card, there is a channel thumbnail for "Programmers are a...", a "Subscribe" button, and social media sharing icons for "25K", "Share", and "...". To the right of the channel information, there are filters for "All", "From Programmers are also h...", "Python", and a right-pointing arrow. The very bottom of the player shows a horizontal bar with the text "Interview with a Senior Python :".

"Mushin lurning"

Interview with a Postdoc, Junior Python Developer

Programmers are a... 154K subscribers

Subscribe

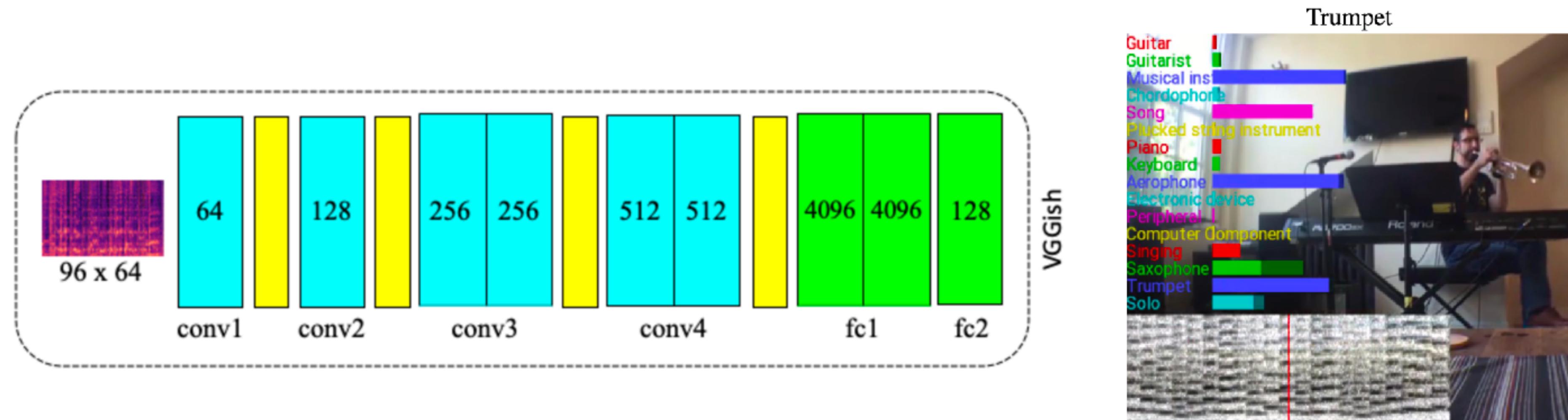
25K Share ...

All From Programmers are also h... Python >

Interview with a Senior Python :

"Audio semantic" models

- **Convolutional neural networks (CNNs)** applied to short [~1 s] spectrograms to generate text labels of audio data (e.g., 'babyCry', 'dogBark')
- Trained on **video-audio correspondence** to learn **the second-order acoustics** ("this kind of spectrogram often goes with that video").



VGGish: Hershey et al., 2017, ICASSP

A possible mid/high-level representation of sounds

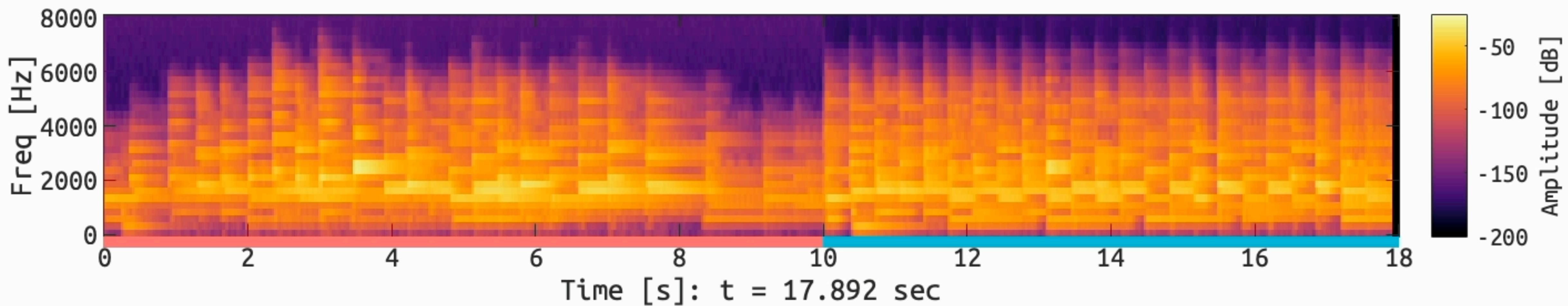
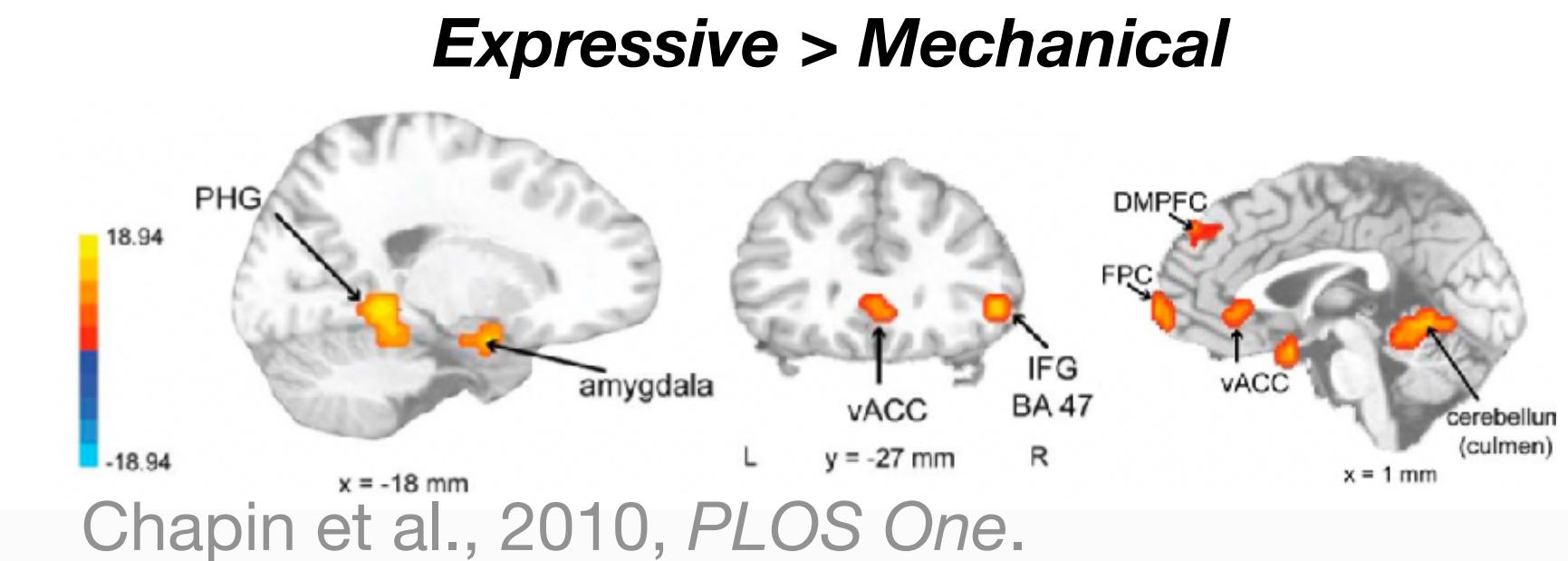


**But really? Isn't it
just timbre?**

Example: Expressive vs. Mechanical

Chapin et al., 2010, PLOS ONE.

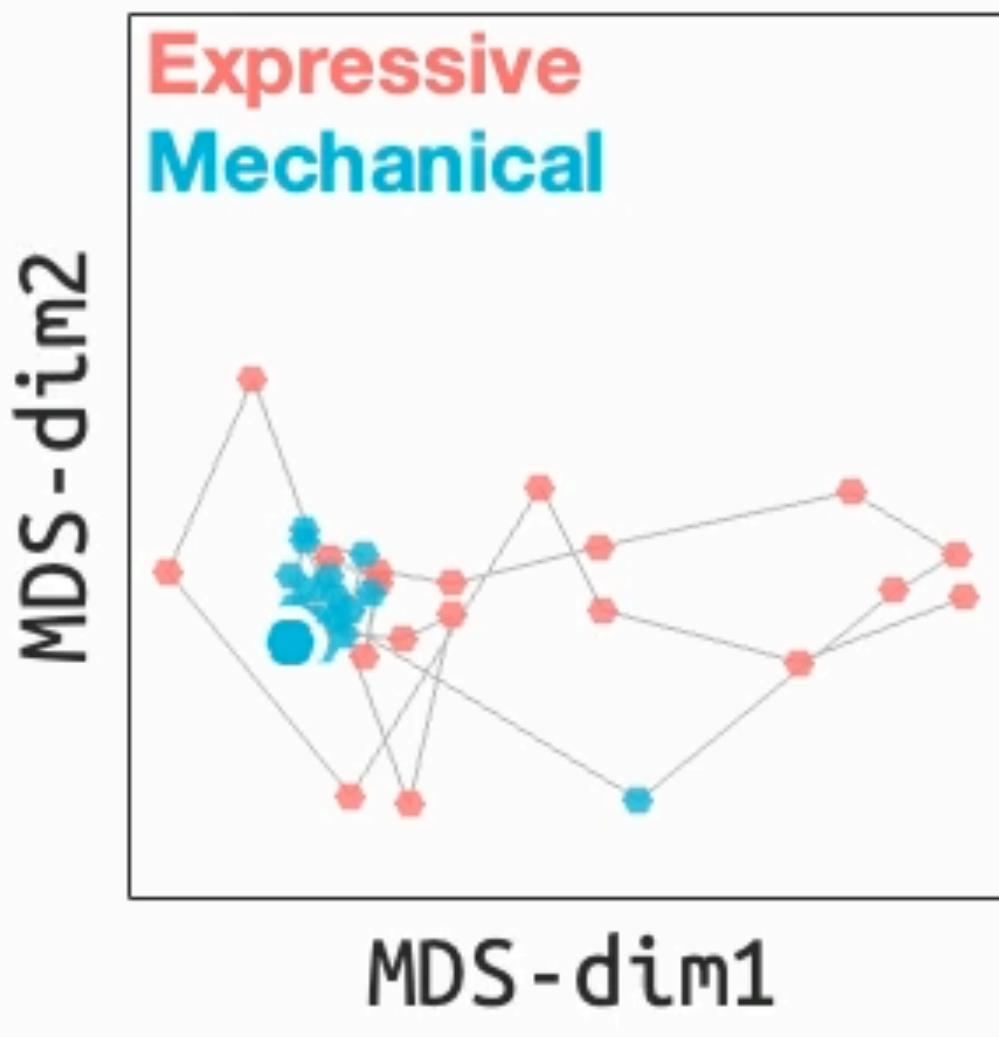
[EXPRESSIVE]: "Frédéric Chopin's Etude in E major, Op.10, No. 3 was performed by an undergraduate piano major (female, 22 years old) on a Kawai CA 950 digital piano"



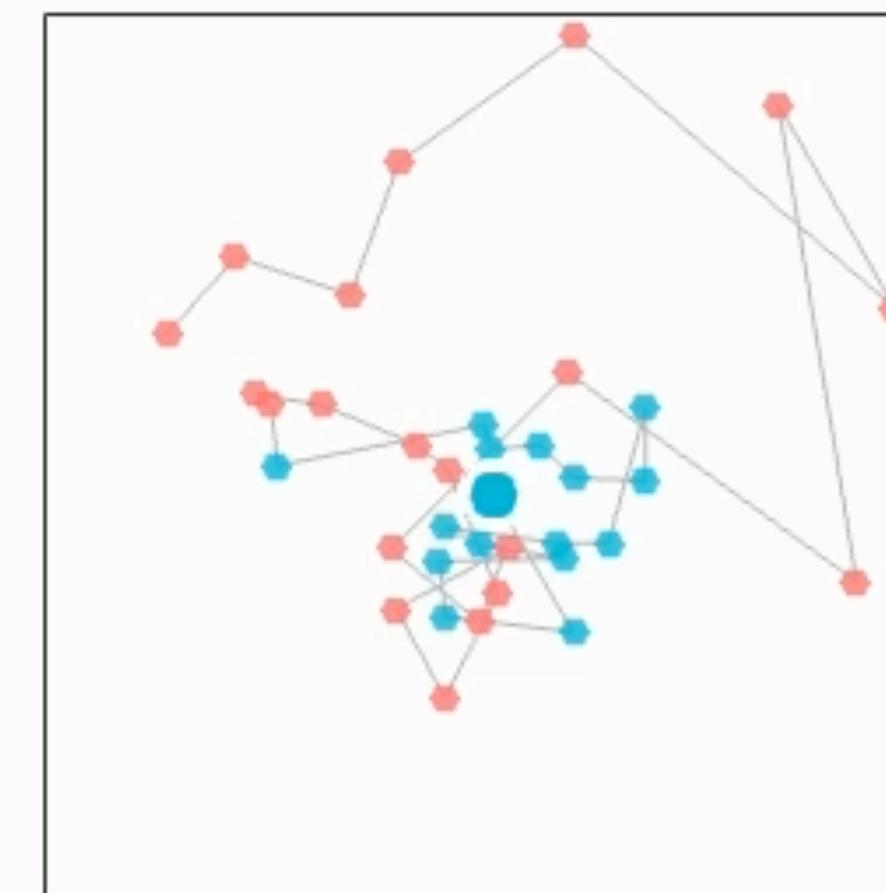
[MECHANICAL]: "The MIDI (Musical Instrument Digital Interface) onset velocity (key pressure) of each note (correlating with sound level) was set to 64 (range 0–128), and pedal information was eliminated."

Example: Expressive vs. Mechanical VGGish representations at various layers (from 1 to 23)

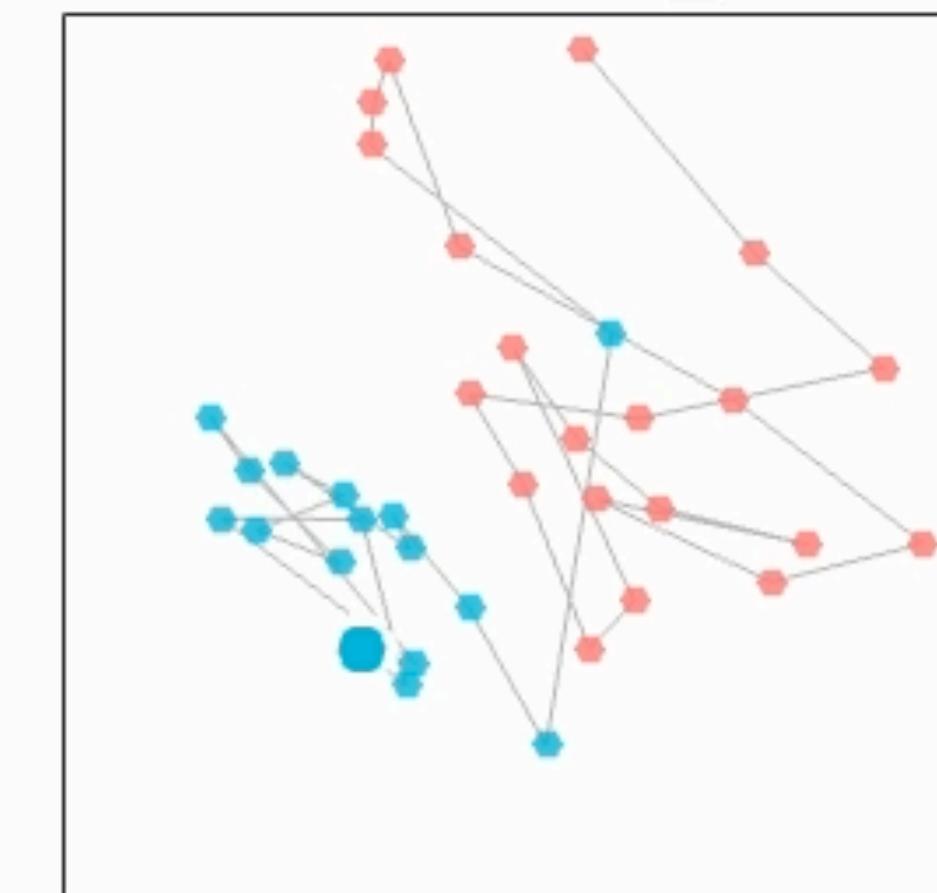
L01: InputBatch



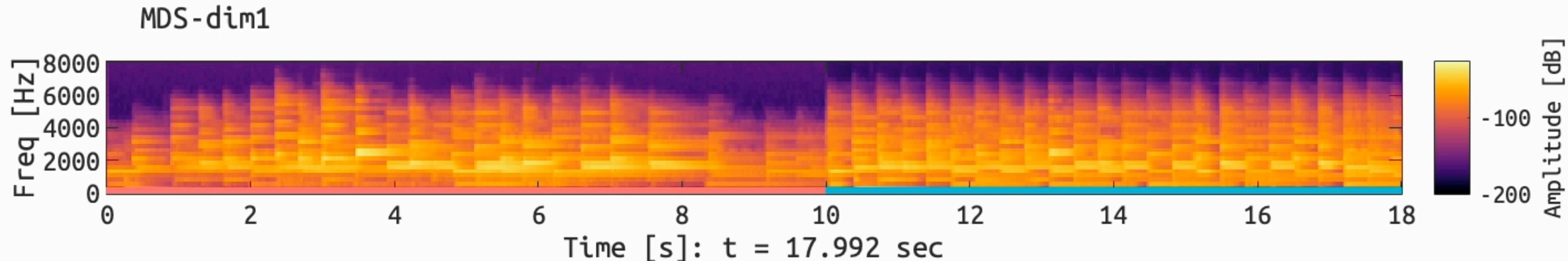
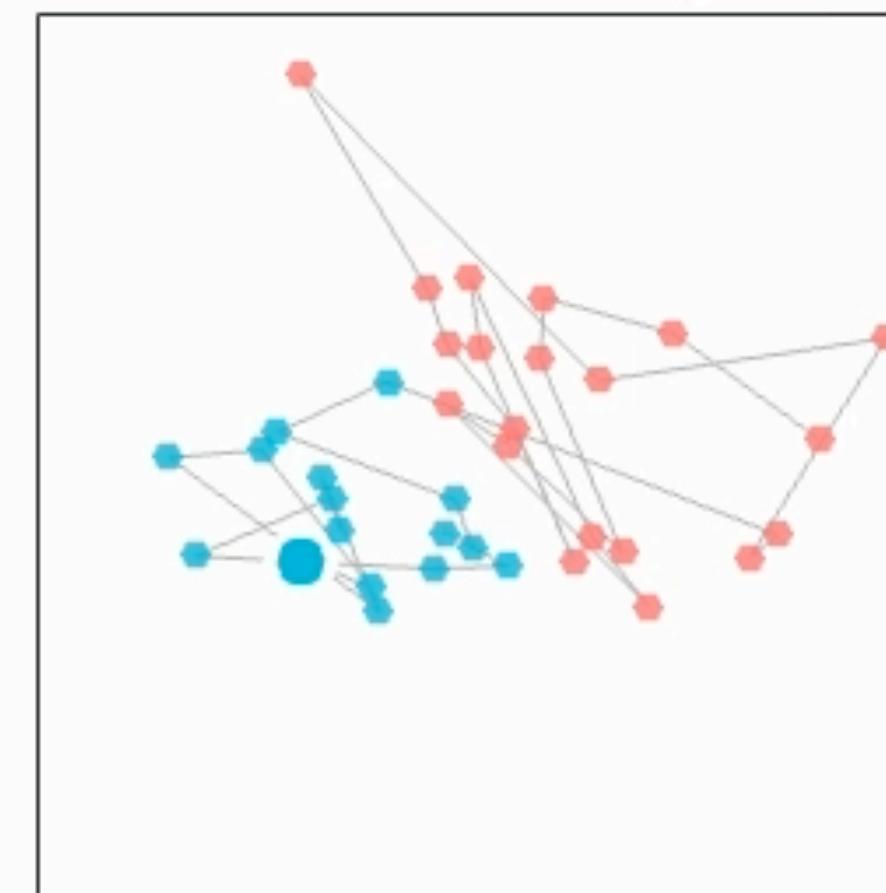
L02: conv1



L18: fc1_1

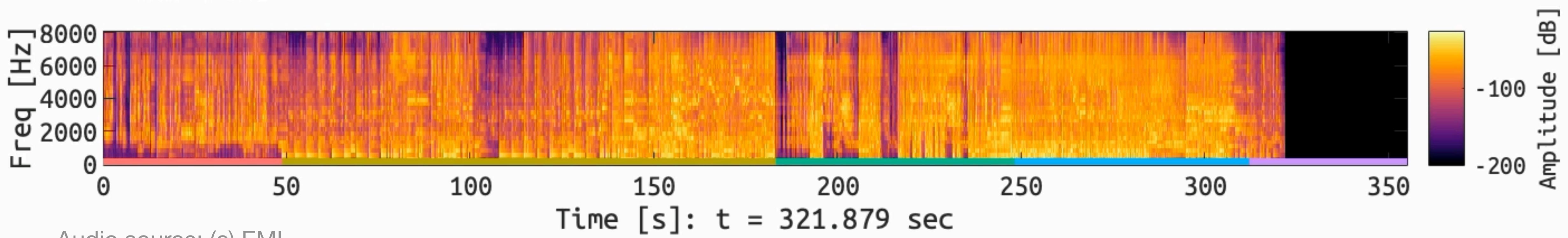
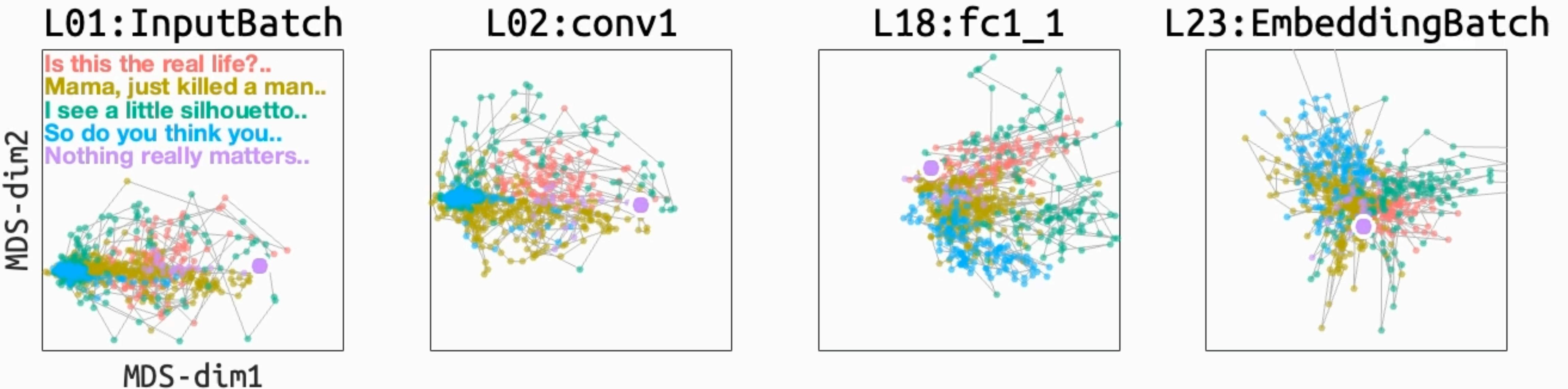


L23: EmbeddingBatch



Example: Queen. (1975). Bohemian Rhapsody

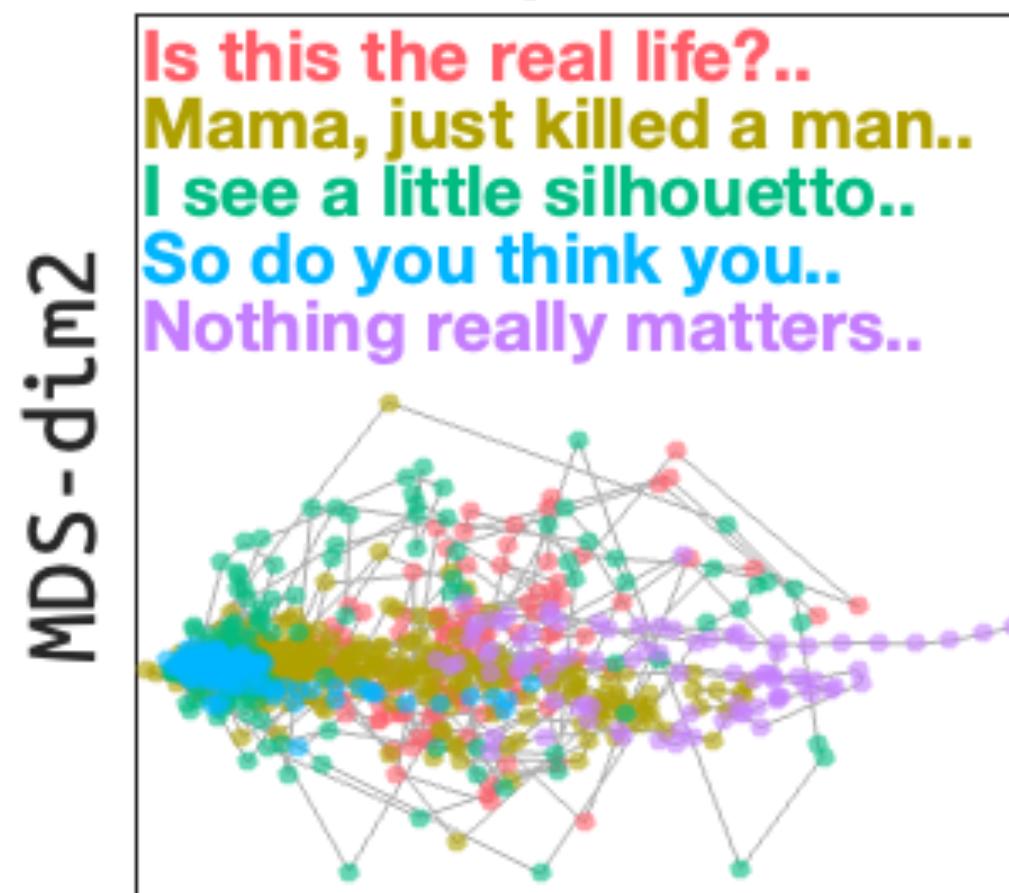
VGGish representations at various layers (from 1 to 23)



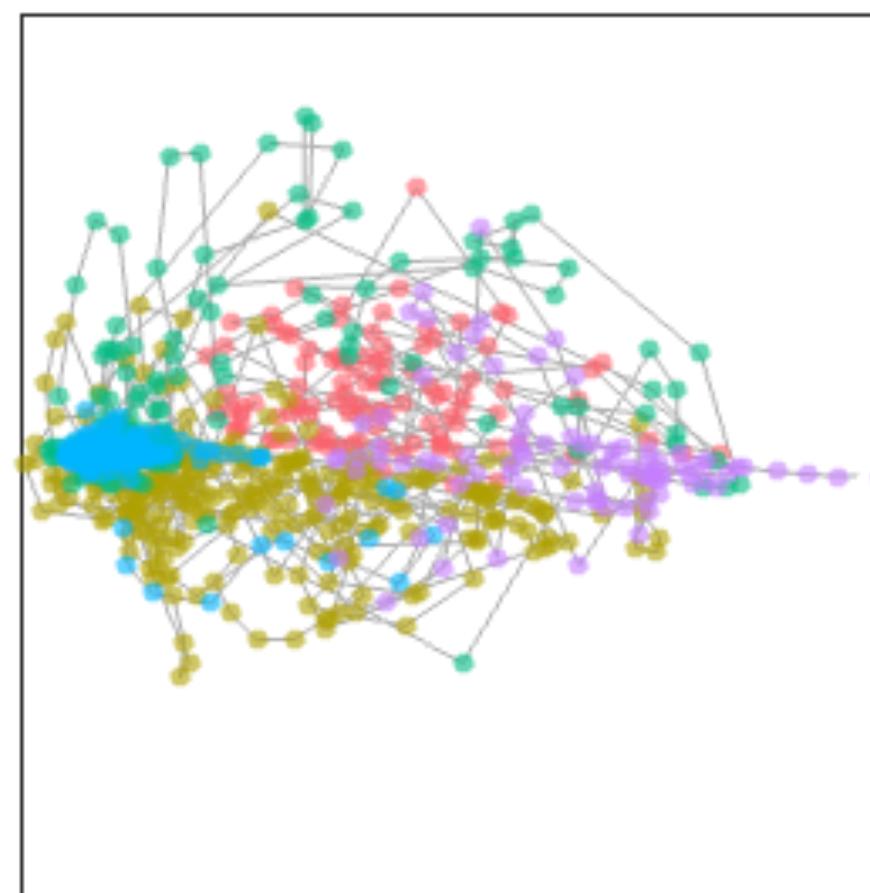
Example: Queen. (1975). Bohemian Rhapsody

VGGish representations at various layers (from 1 to 23)

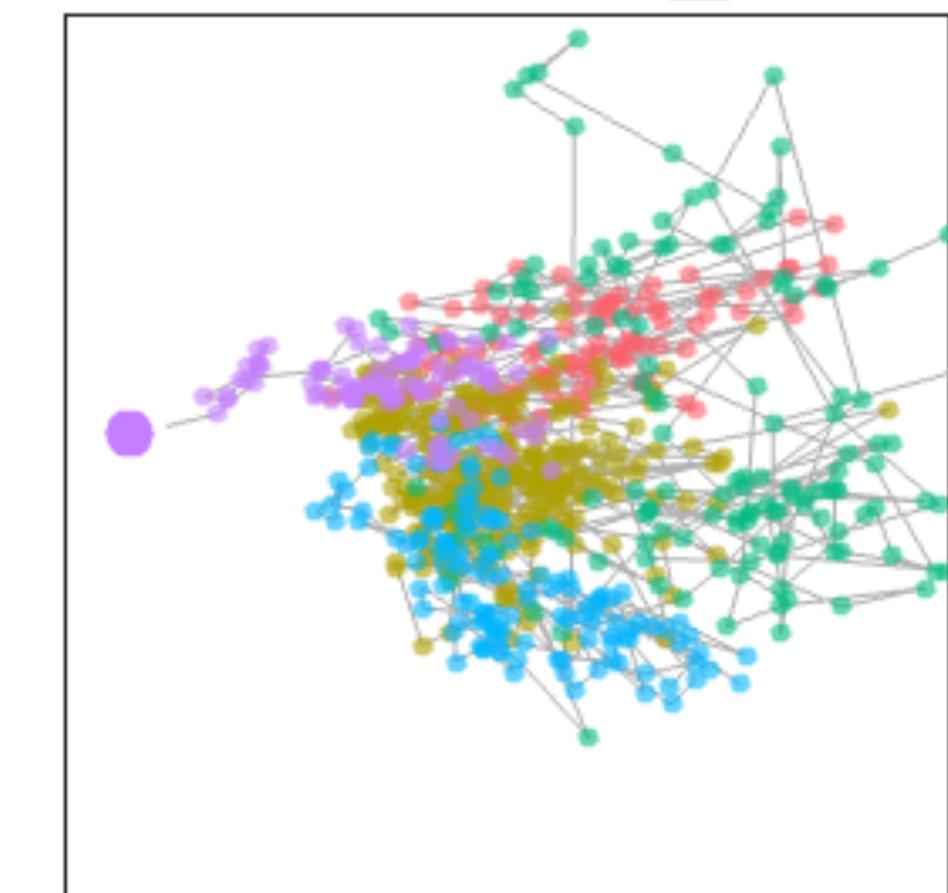
L01:InputBatch



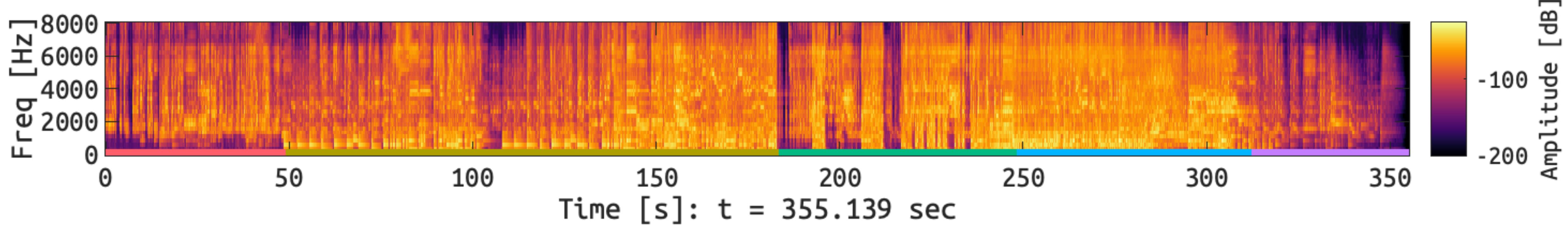
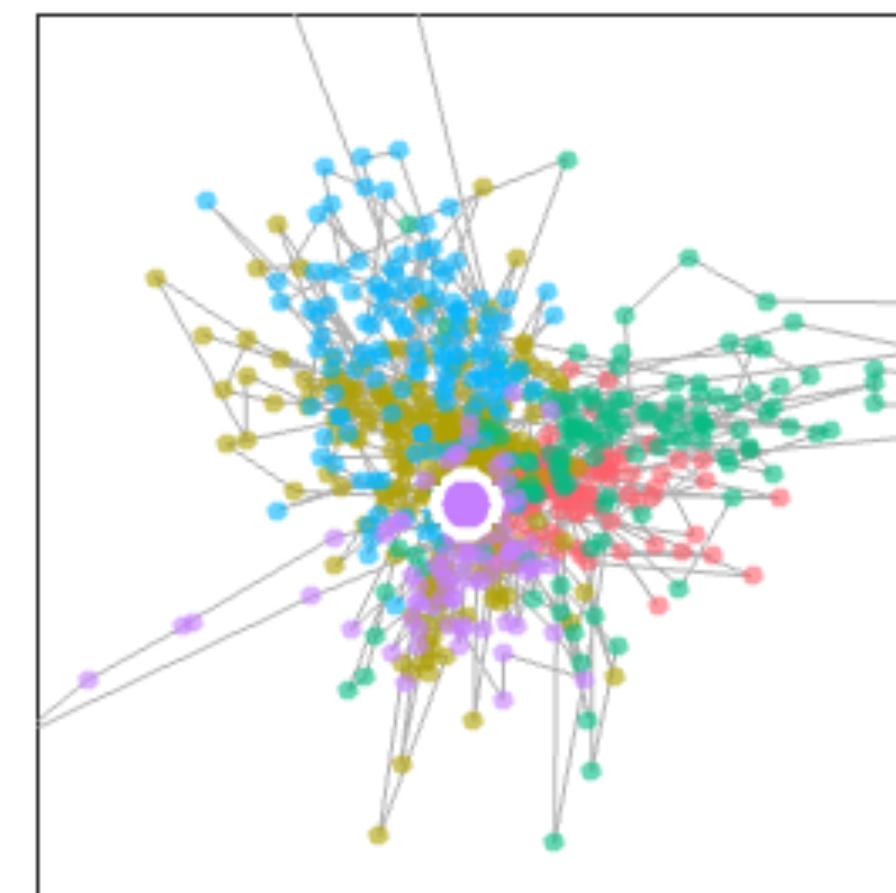
L02:conv1



L18:fc1_1



L23:EmbeddingBatch



CNN embedding for music emotion *recognition*

Potentially mid/high-level representation of music signal

An "deep audio semantic" model called "VGGish"

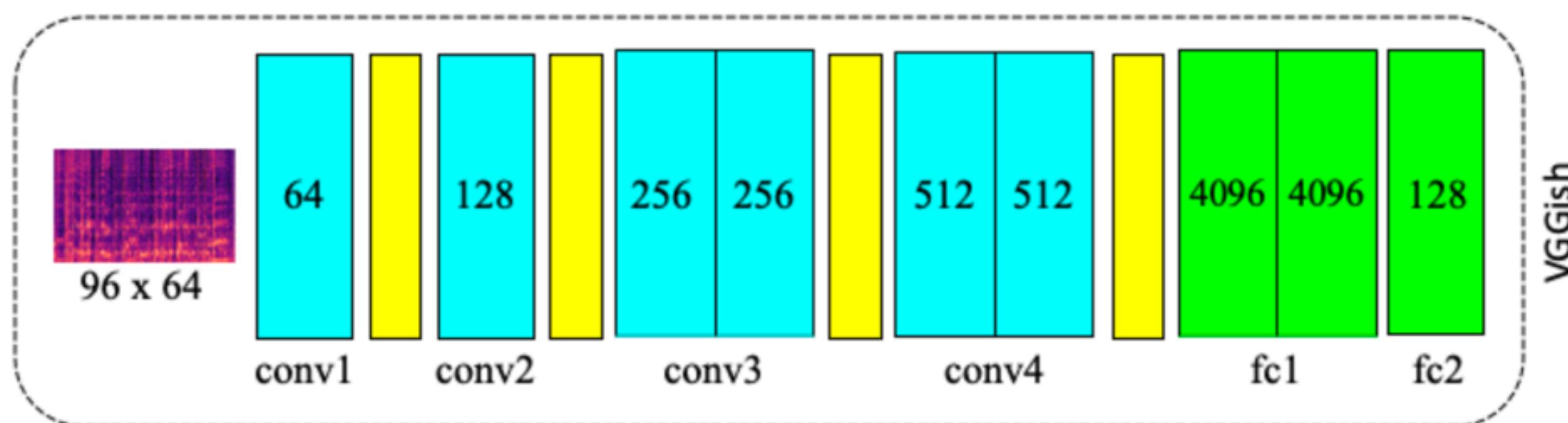
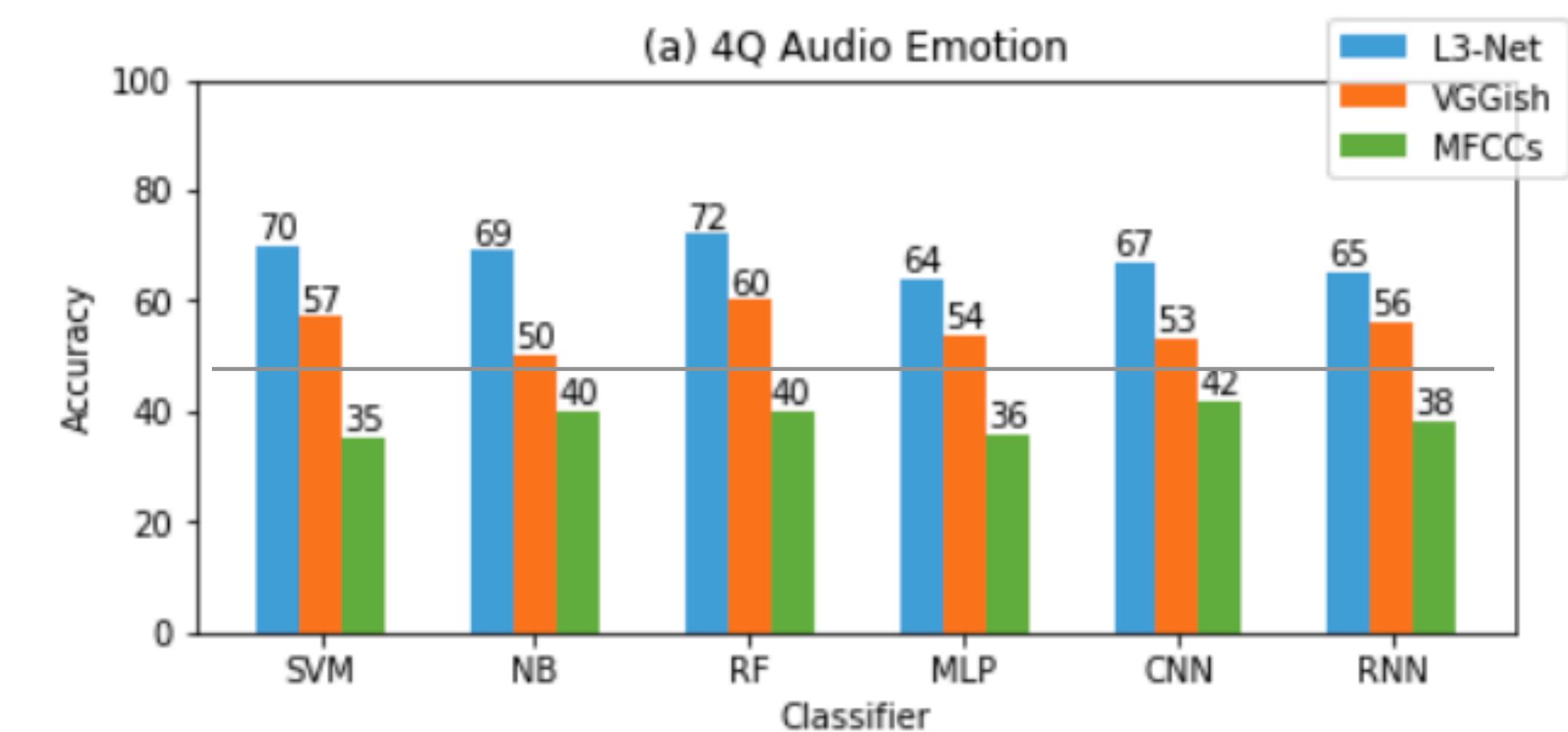


Diagram from Koh & Dubnov, 2021, ACA

4Q Audio Emotion Dataset: 255 music clips (30 s) for Arousal-Valence quadrants



Koh & Dubnov, 2021, ACA

Deep audio semantic models carry more information related to expressed emotions than a traditional audio descriptor.



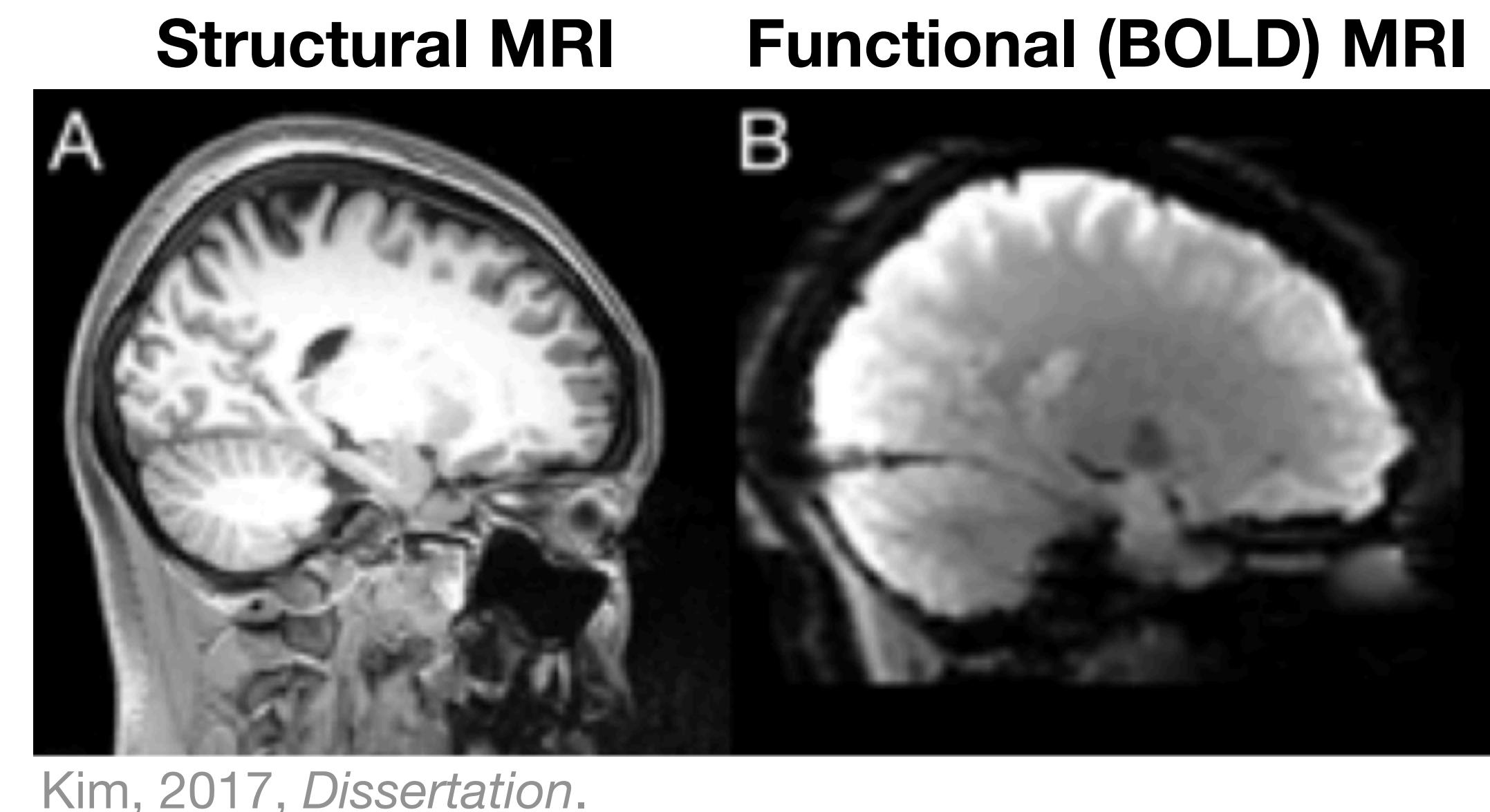
How do we measure brain activity?



Functional magnetic resonance imaging

Still the state-of-the-art non-invasive functional imaging

- **Hardware:** the same MRI at hospitals (but with a different program)
- **Mechanism:** Blood-oxygenation-level-dependent (BOLD) effect
- **Strengths:** the highest spatial (3-mm-iso) and temporal resolution (1 sec) than any other non-invasive imaging (e.g., PET)
- **Weaknesses:** indirect neural activity (unlike M/EEG, ECoG), ~120 dB acoustic scanning noise (but there are active noise cancelling headphones for MRI)



<https://www.irasutoya.com/>



How do we measure emotions?



Felt vs. perceived emotions

We just ask people how they feel...

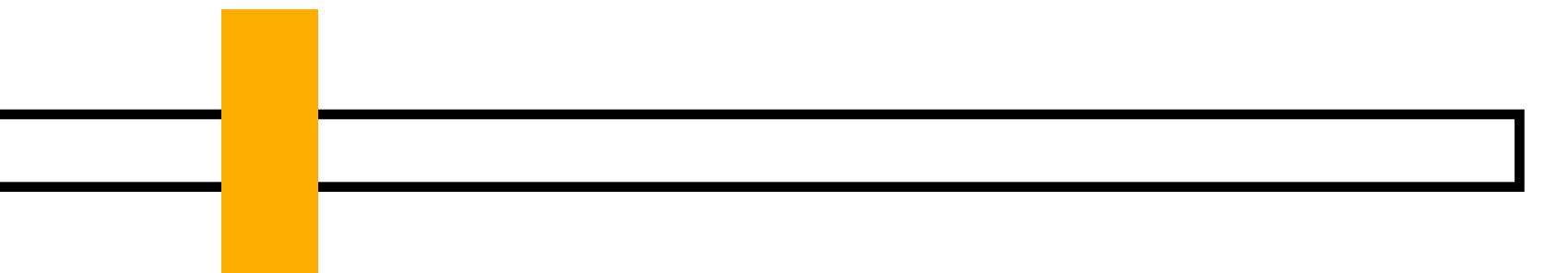
Rating with a slider



<https://www.irasutoya.com/>

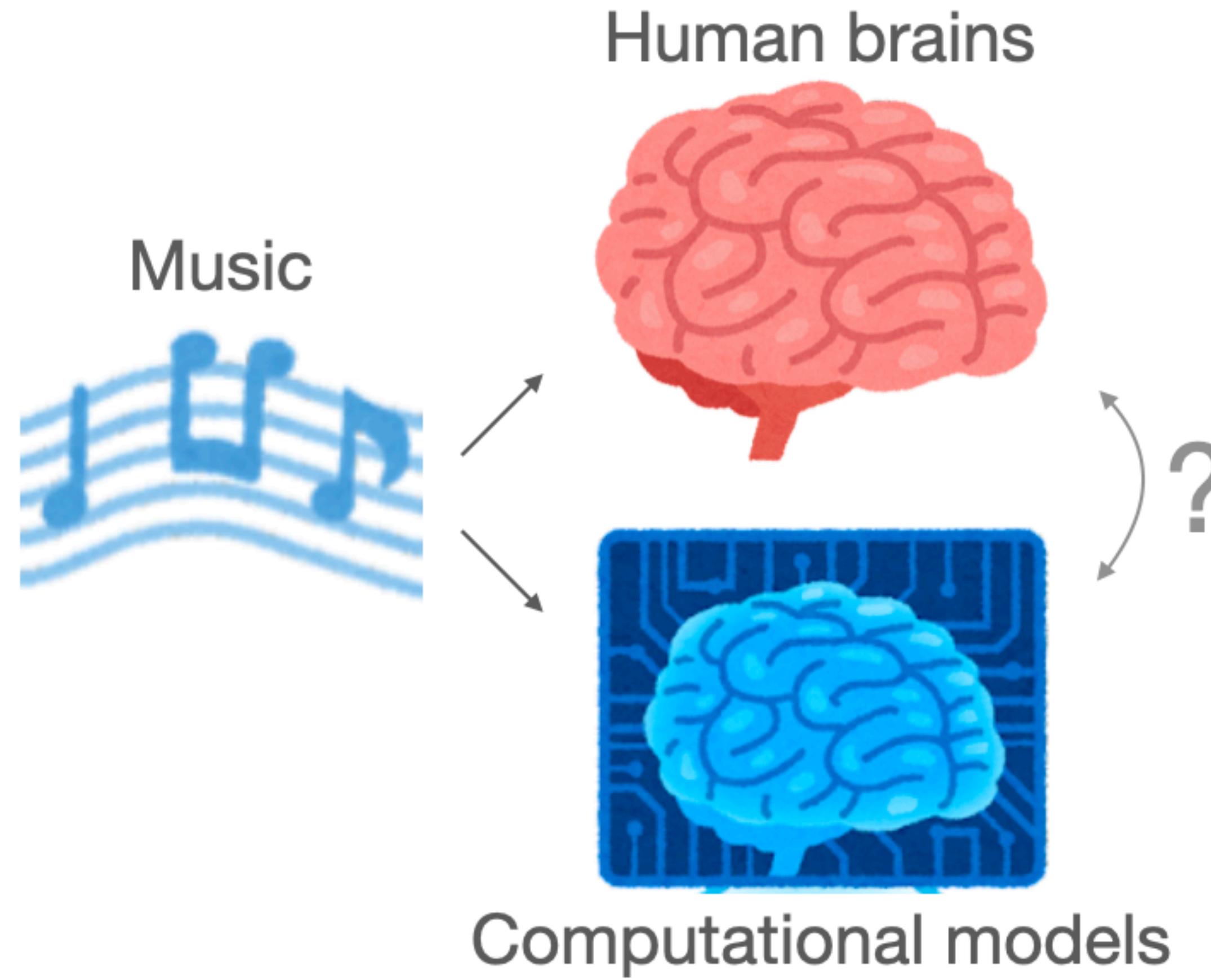
How intensively do you feel SAD/HAPPY right now?

Not at all



Extremely

Research Questions



- **Q1:** How are *felt emotions* and **musical enjoyment** associated with neural activity over time?
- **Q2:** Would **increasingly abstract representations of music** in different layers of the CNN be encoded along the cortical gradient axis of abstraction?
- **Q3:** How do **layer-specific CNN embeddings** predict human behavioral ratings of musical emotions?



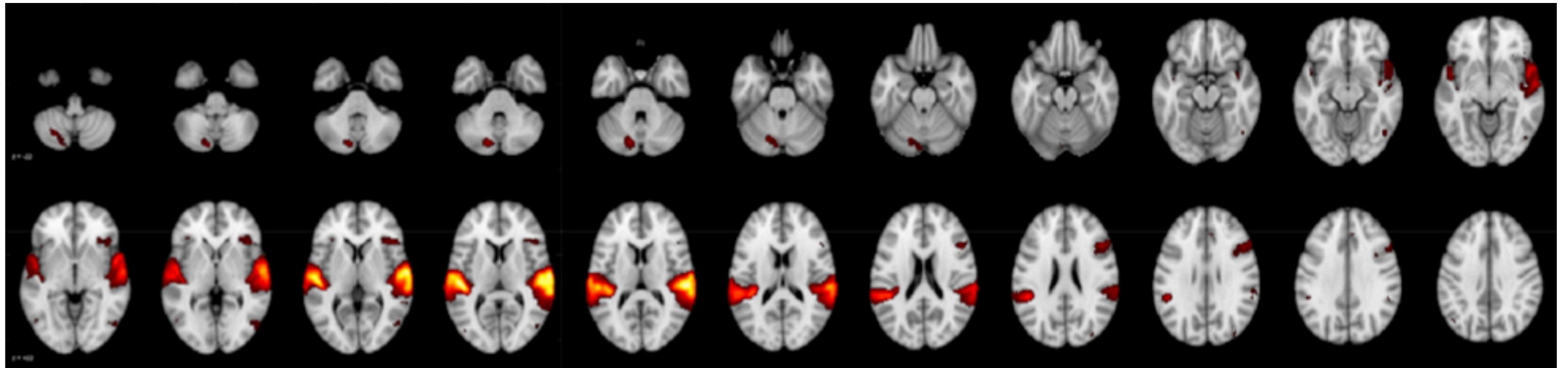
Methods

Original study

Sachs et al., 2020, *NeuroImage*.



Inter-subject correlation during a "sad" piece of music: $r \sim [0, 0.16]$, cluster- $P < 0.05$

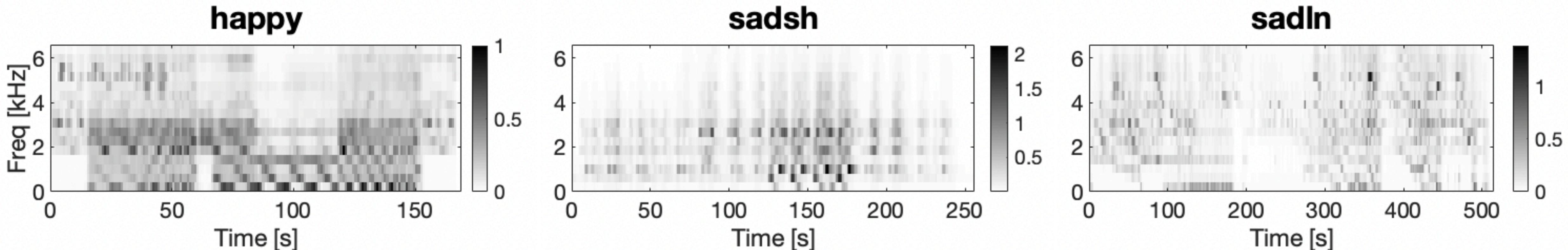


Sachs et al., 2020, *NeuroImage*.

Stimuli

Sachs et al., 2020, *NeuroImage*.

- **Happy** [2 min 48 sec]: Lullatone's "Race against the Sunset"
- **Sad-short** [4 min 16 sec]: Olafur Arnalds's "Frysta"
- **Sad-long** [8 min 35 sec]: Michael Kamen's "Discovery of the Camp"



Kim et al., *In prep.*

Participants & protocol

Sachs et al., 2020, *NeuroImage*.

- N = 40 (21 female, mean age = 24.1 ± 6.24 from LA)
 - Unfamiliar with 3 stimuli and reported "intended" emotions from 60-s excerpts

Passive listening with eyes open

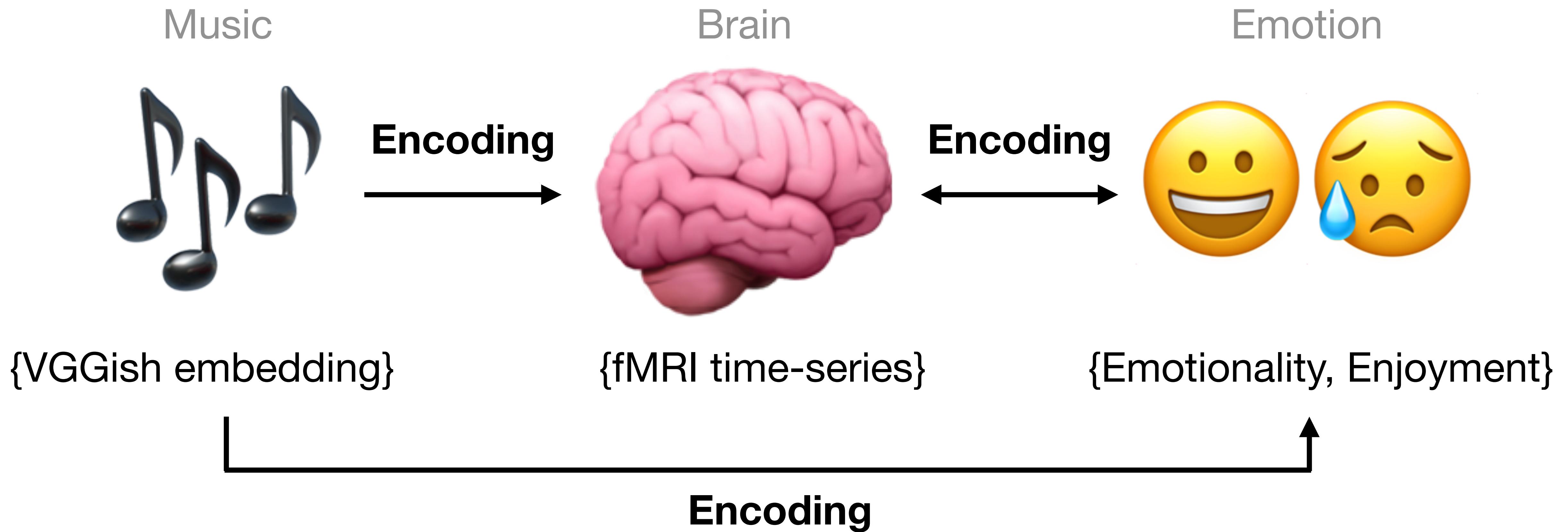


Rating with a slider



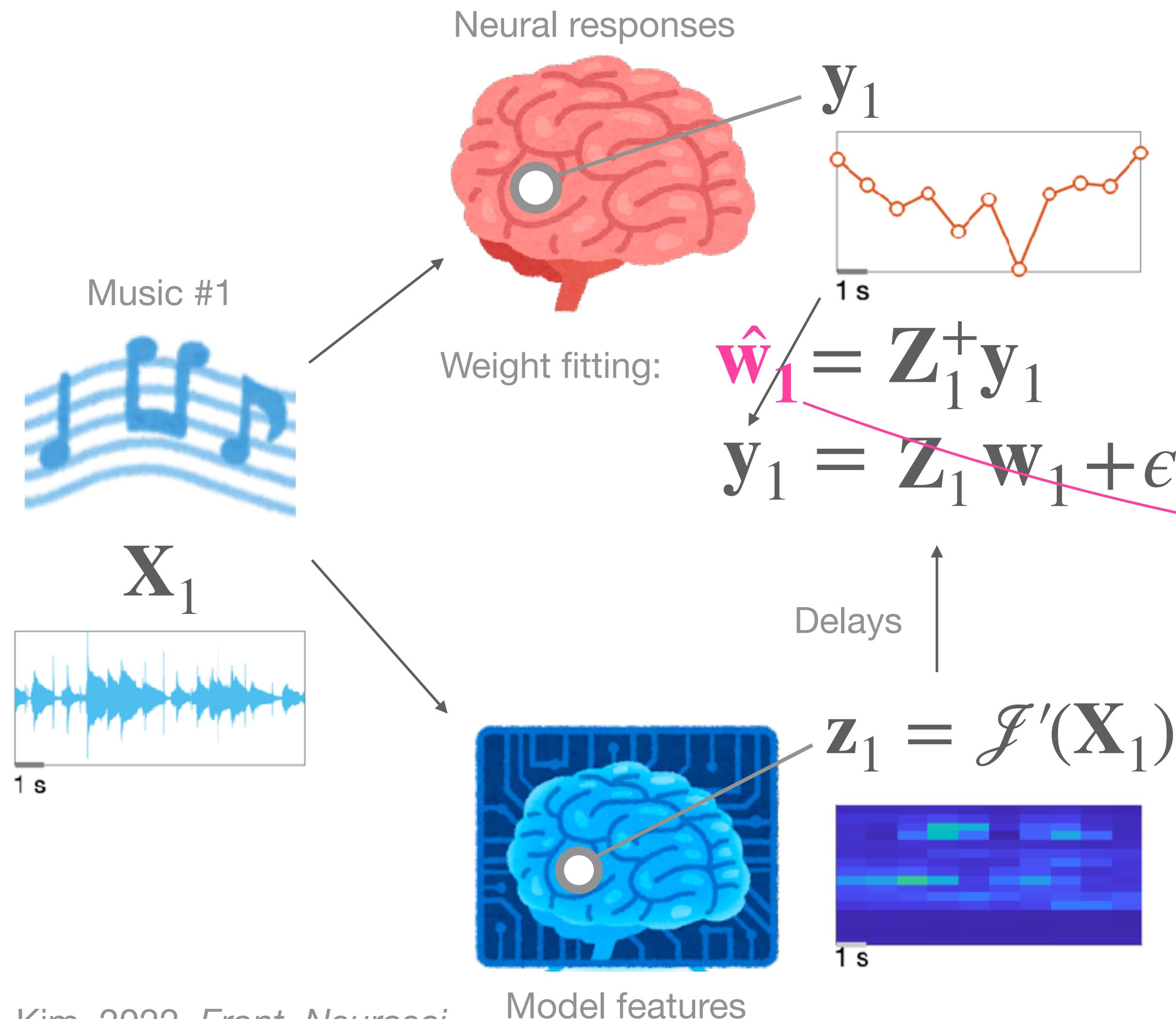
- "The intensity of felt sadness or happiness" (**Emotionality**)
- "The intensity of enjoyment" (**Enjoyment**)

Analysis overview

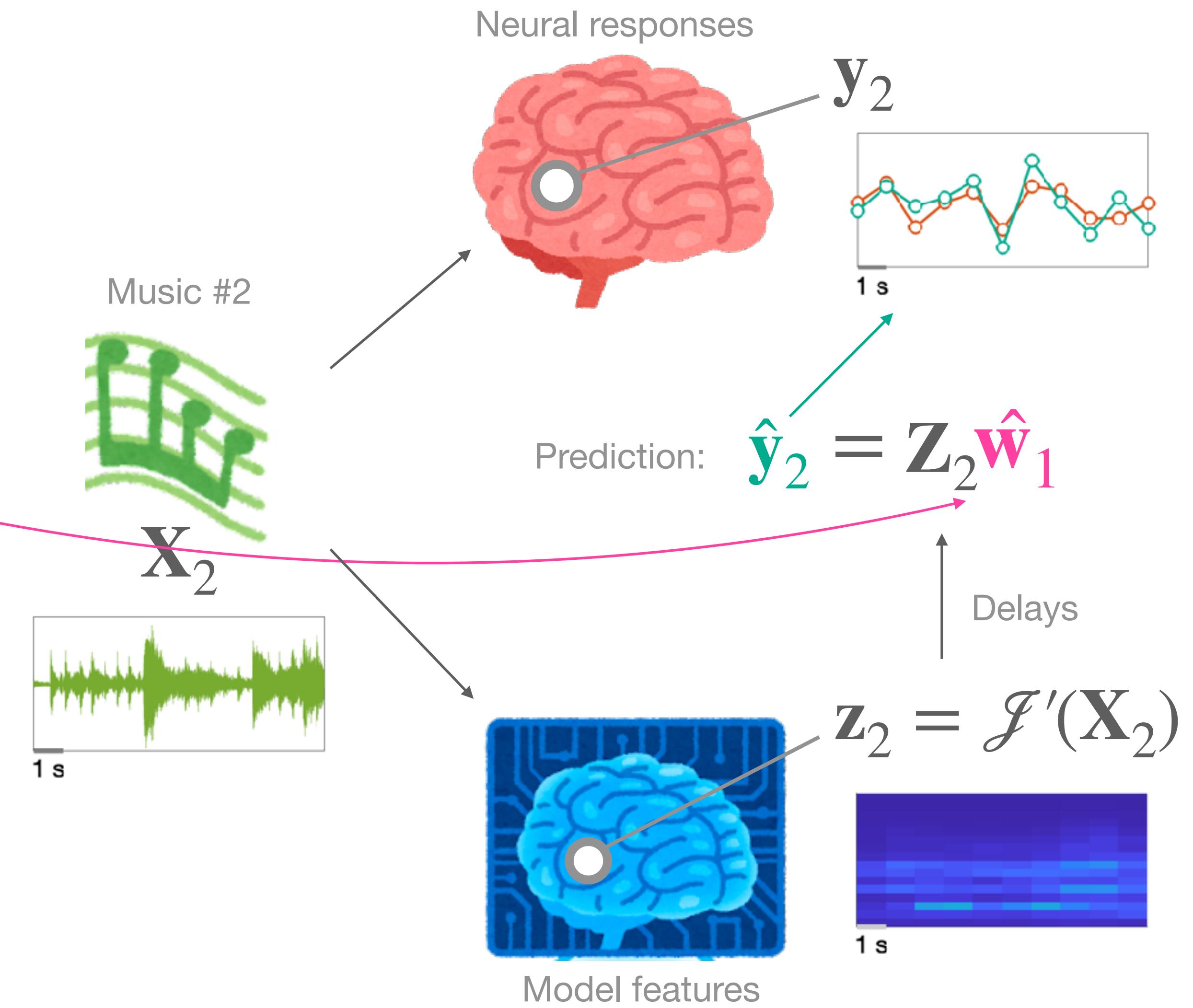


Encoding analysis

Training set

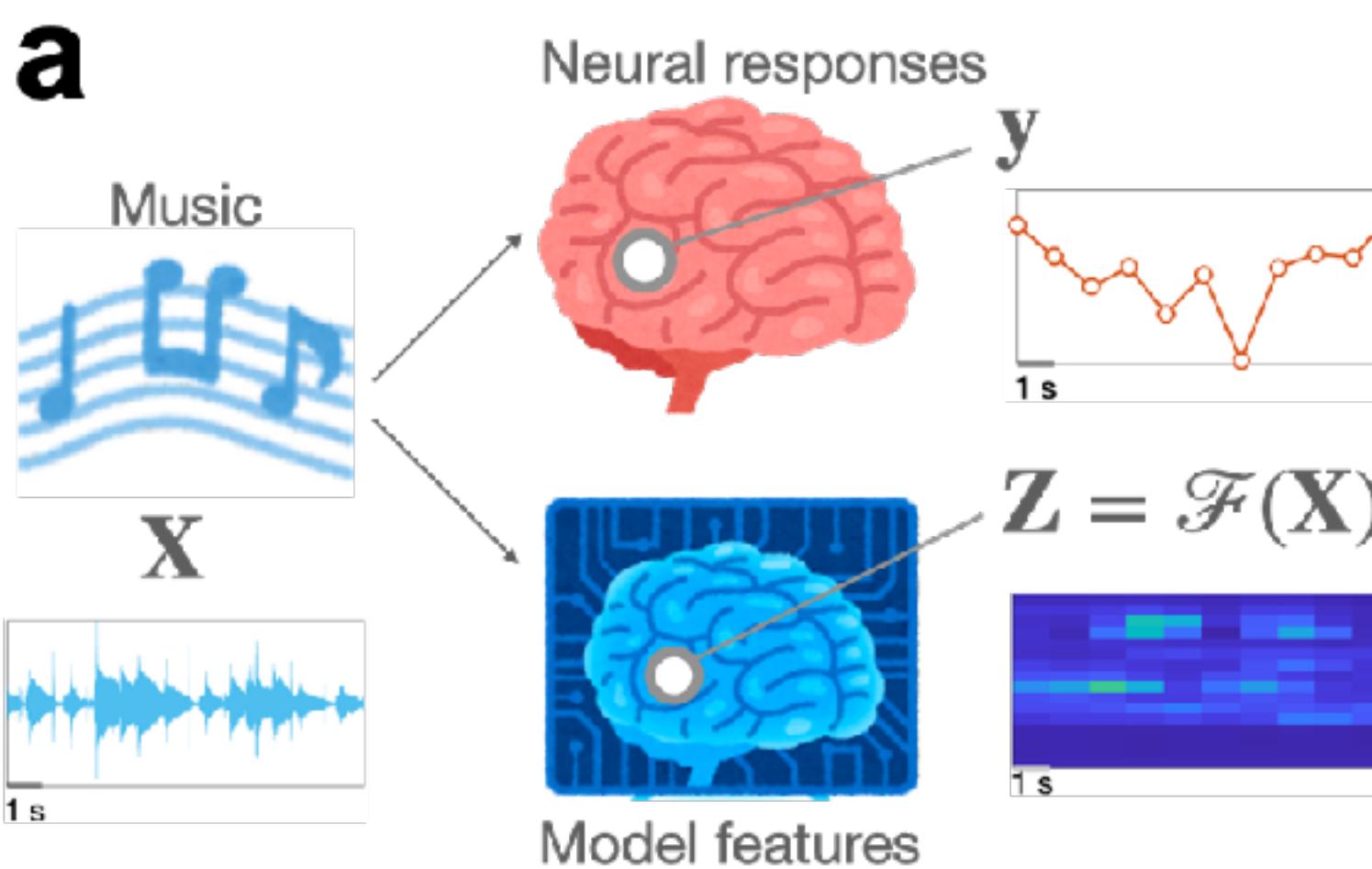


Test set



Layer-specific profile

How do we map VGGish-layer-specific encoding on the cortex?

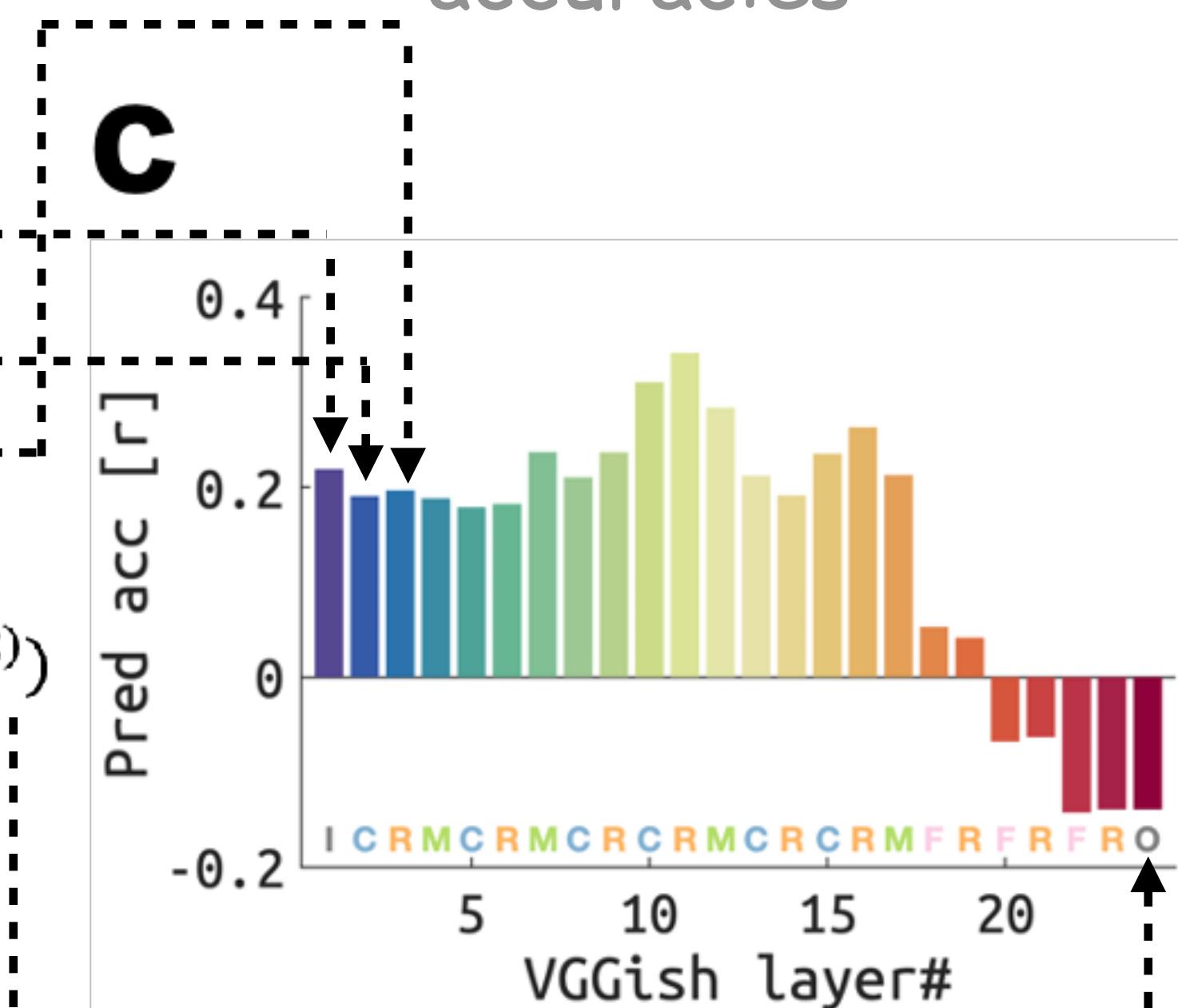


How good the "Layer X" is for this brain area?

b

$$\begin{aligned}y^{(1)} &= \mathcal{F}_1(\mathbf{X}^{(1)})\mathbf{b}_1 + \varepsilon \rightarrow r_1 = \text{corr}(\hat{y}_1^{(2)}, y^{(2)}) \\y^{(1)} &= \mathcal{F}_2(\mathbf{X}^{(1)})\mathbf{b}_2 + \varepsilon \rightarrow r_2 = \text{corr}(\hat{y}_2^{(2)}, y^{(2)}) \\y^{(1)} &= \mathcal{F}_3(\mathbf{X}^{(1)})\mathbf{b}_3 + \varepsilon \rightarrow r_3 = \text{corr}(\hat{y}_3^{(2)}, y^{(2)}) \\&\vdots \\y^{(1)} &= \mathcal{F}_{24}(\mathbf{X}^{(1)})\mathbf{b}_{24} + \varepsilon \rightarrow r_{24} = \text{corr}(\hat{y}_{24}^{(2)}, y^{(2)})\end{aligned}$$

A "profile" of layer-specific prediction accuracies



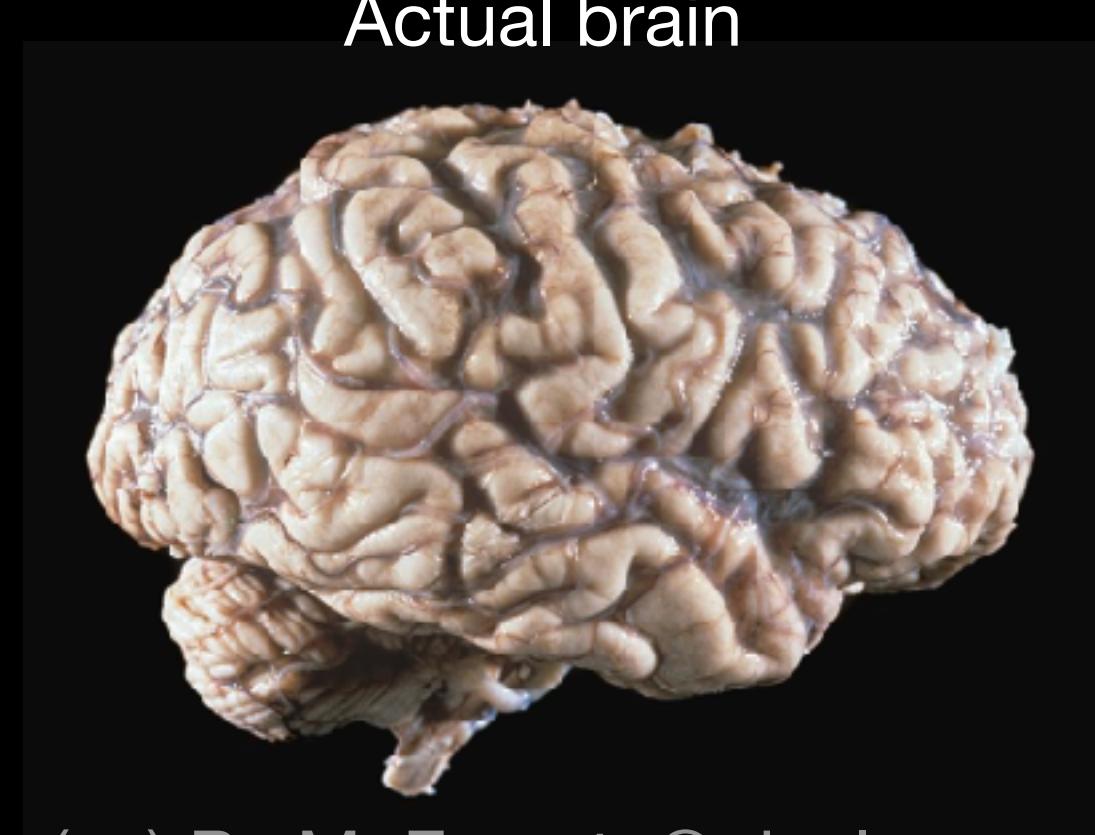


Results

Inflated cortical surface and anatomy

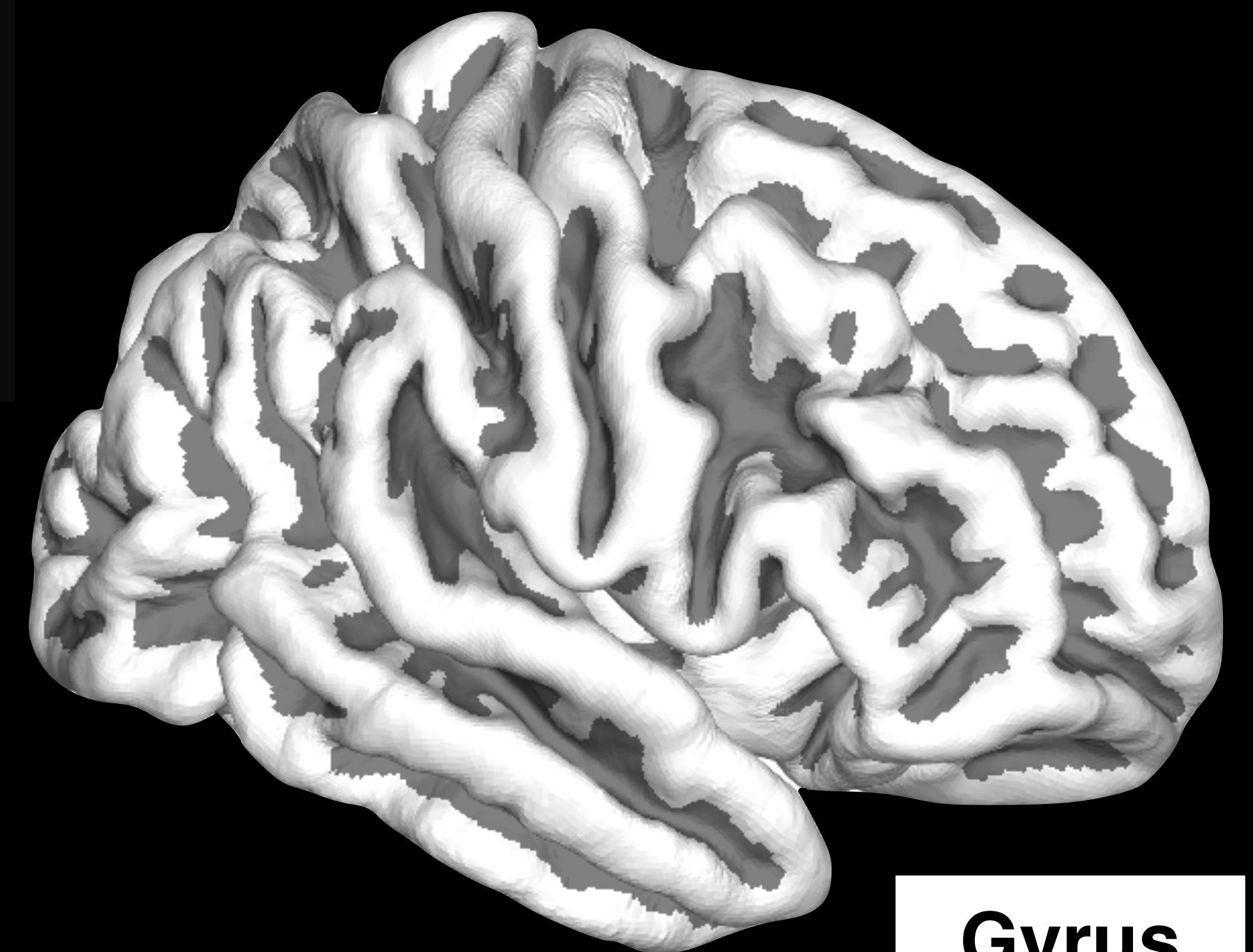
Just to see better the buried parts of the cortex

Actual brain

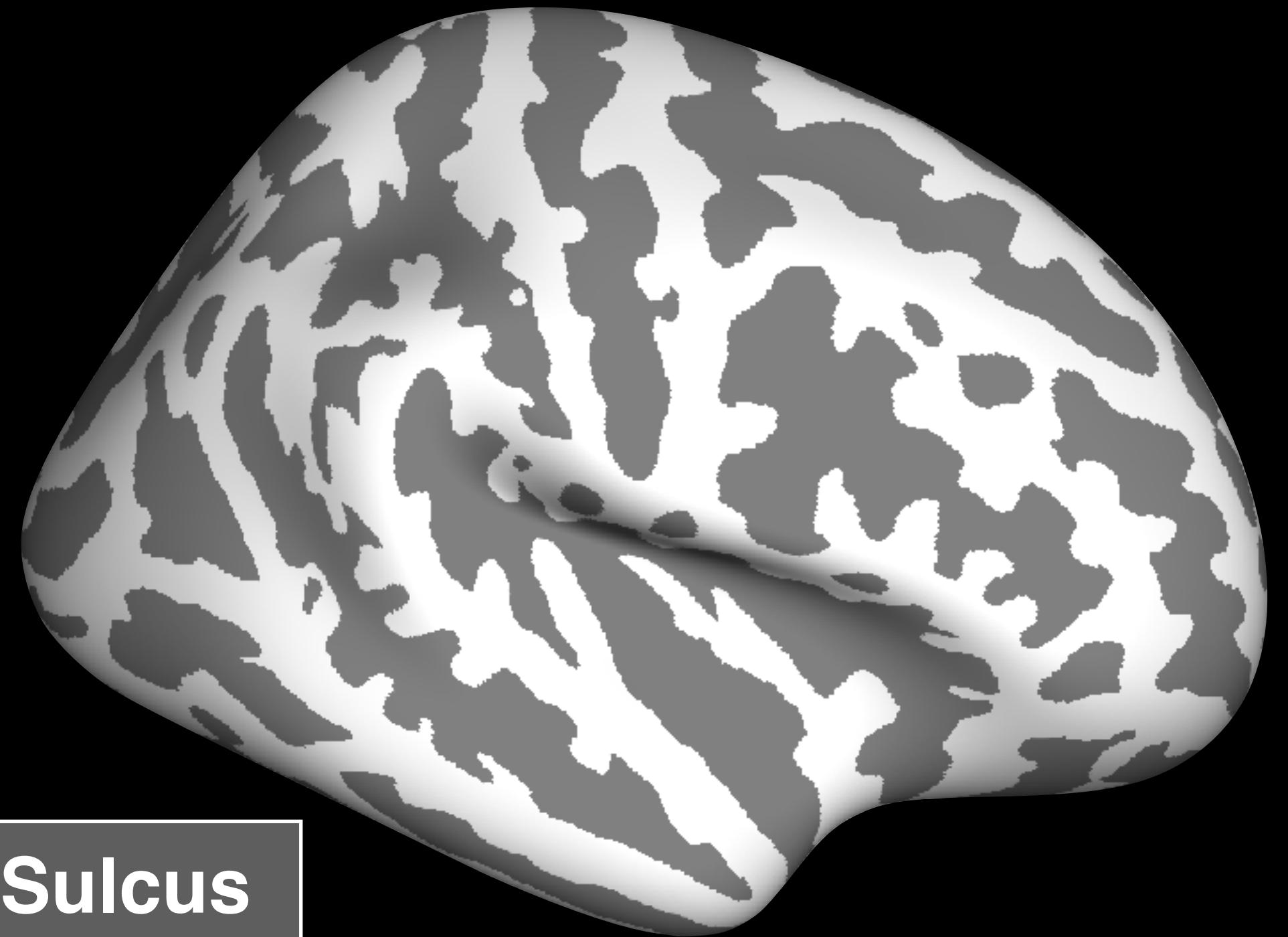


(cc) Pr. M. Forest @pixel.com

"Averaged brain"



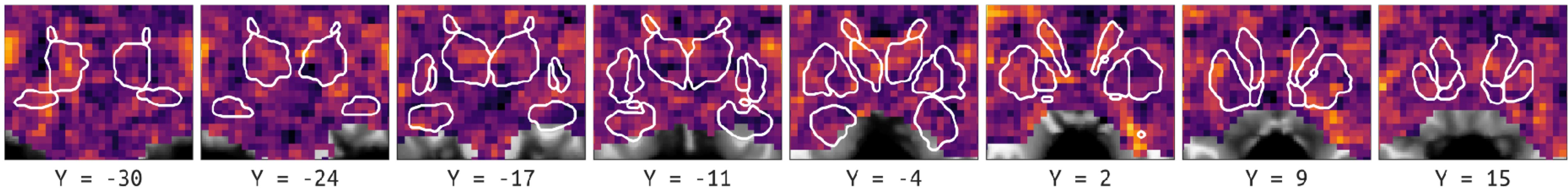
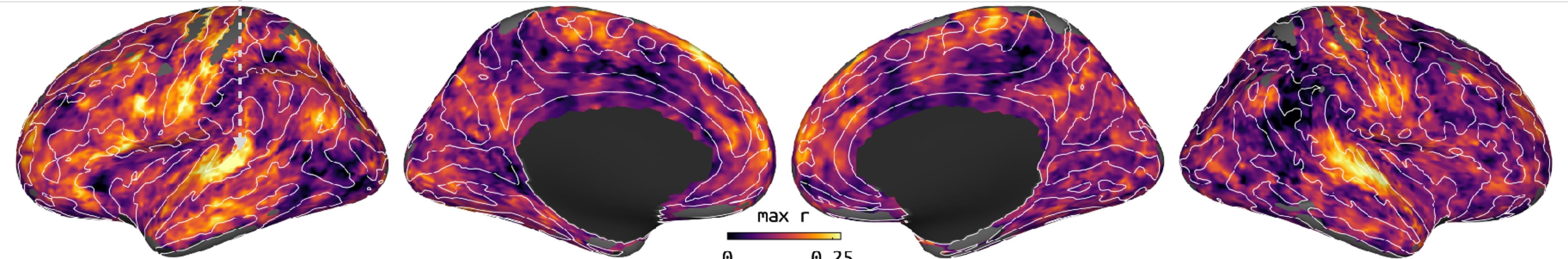
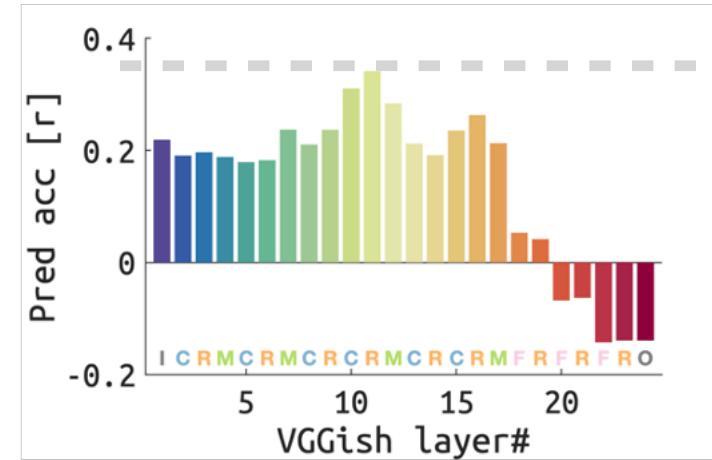
"Inflated brain"



Gyrus

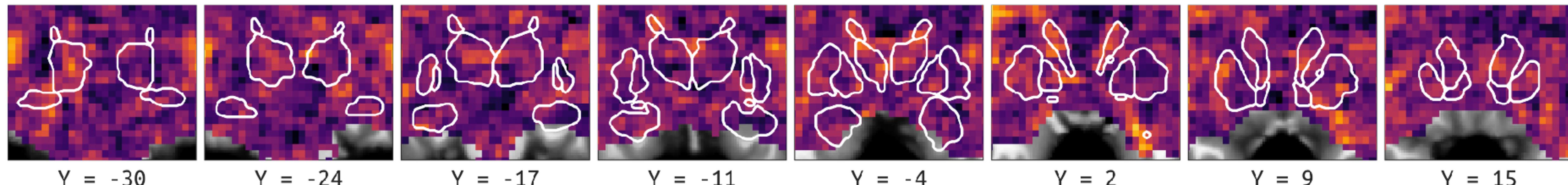
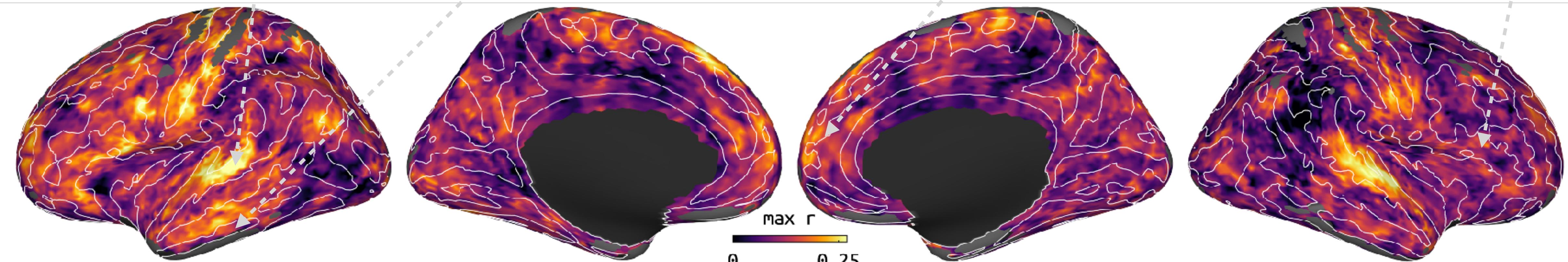
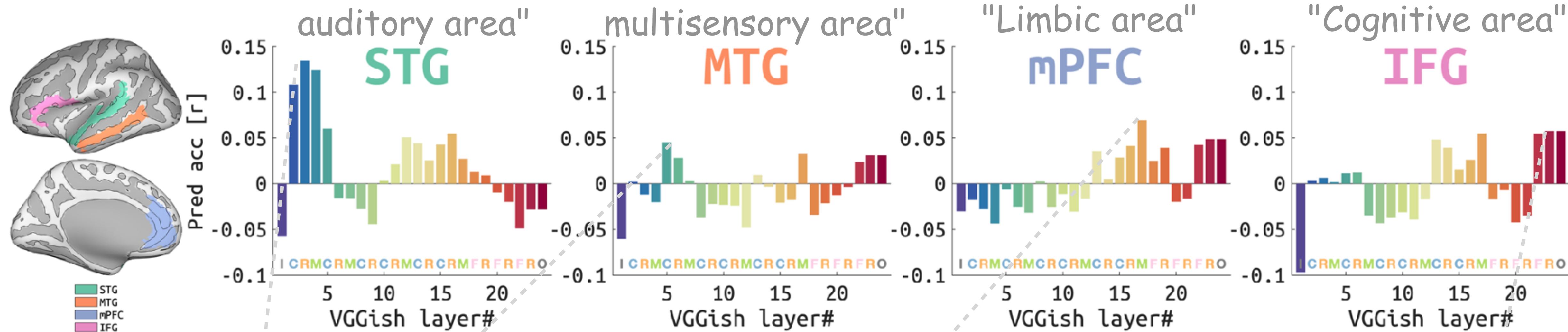
Sulcus

Maximal prediction accuracy ♪♪ →

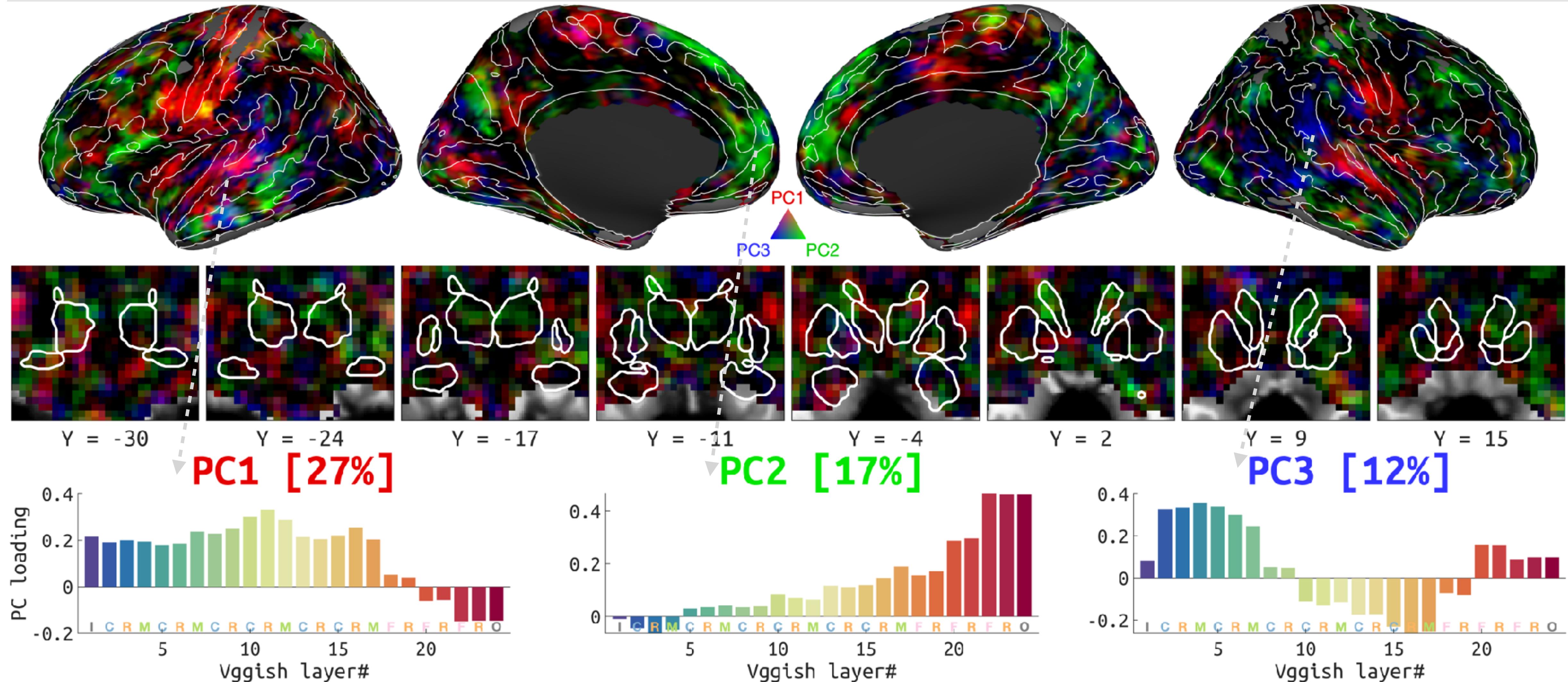
C

*Simplified labels: "Low-level

"High-level

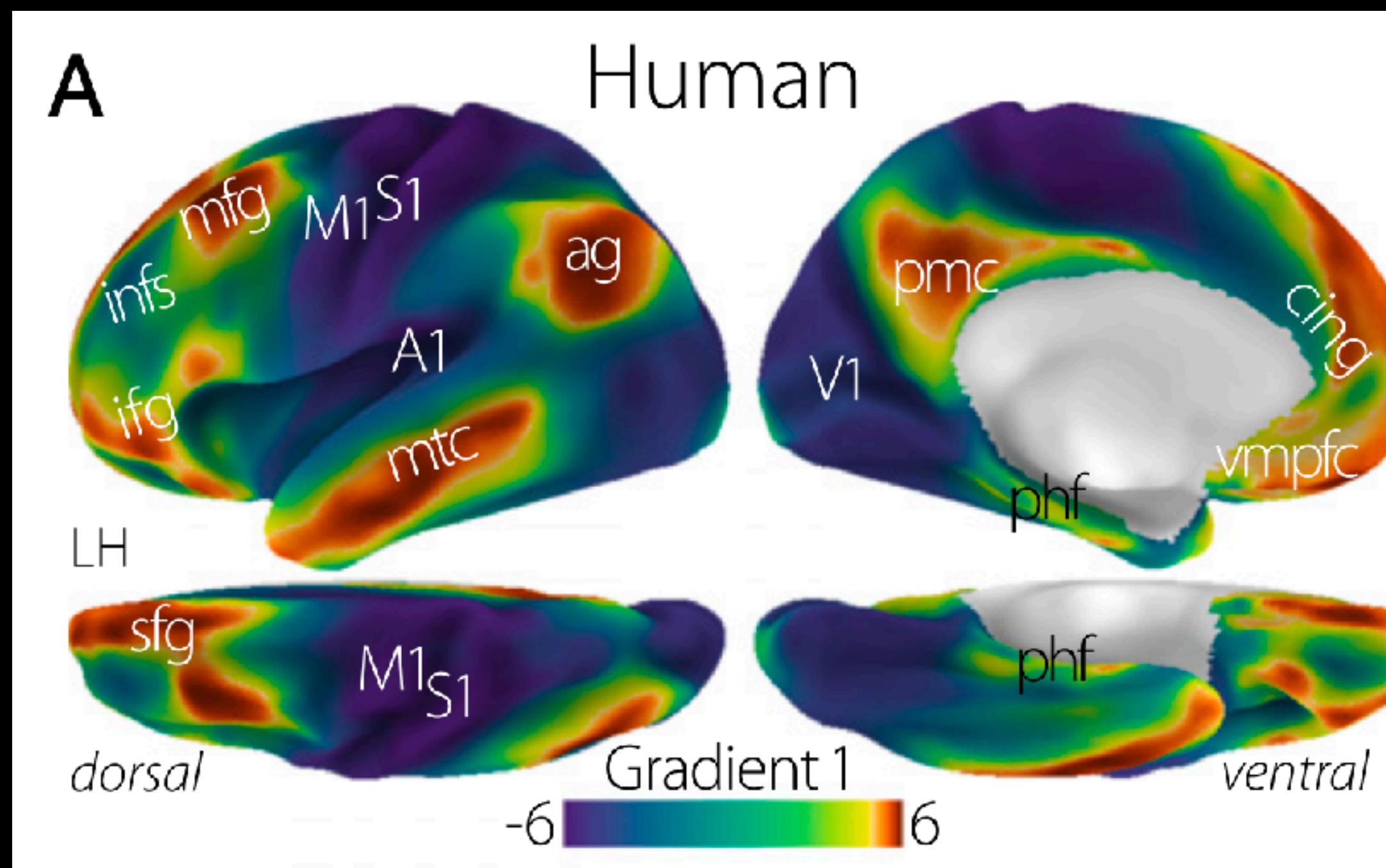


Topography of CNN-layer-specific encoding 🎵→🧠

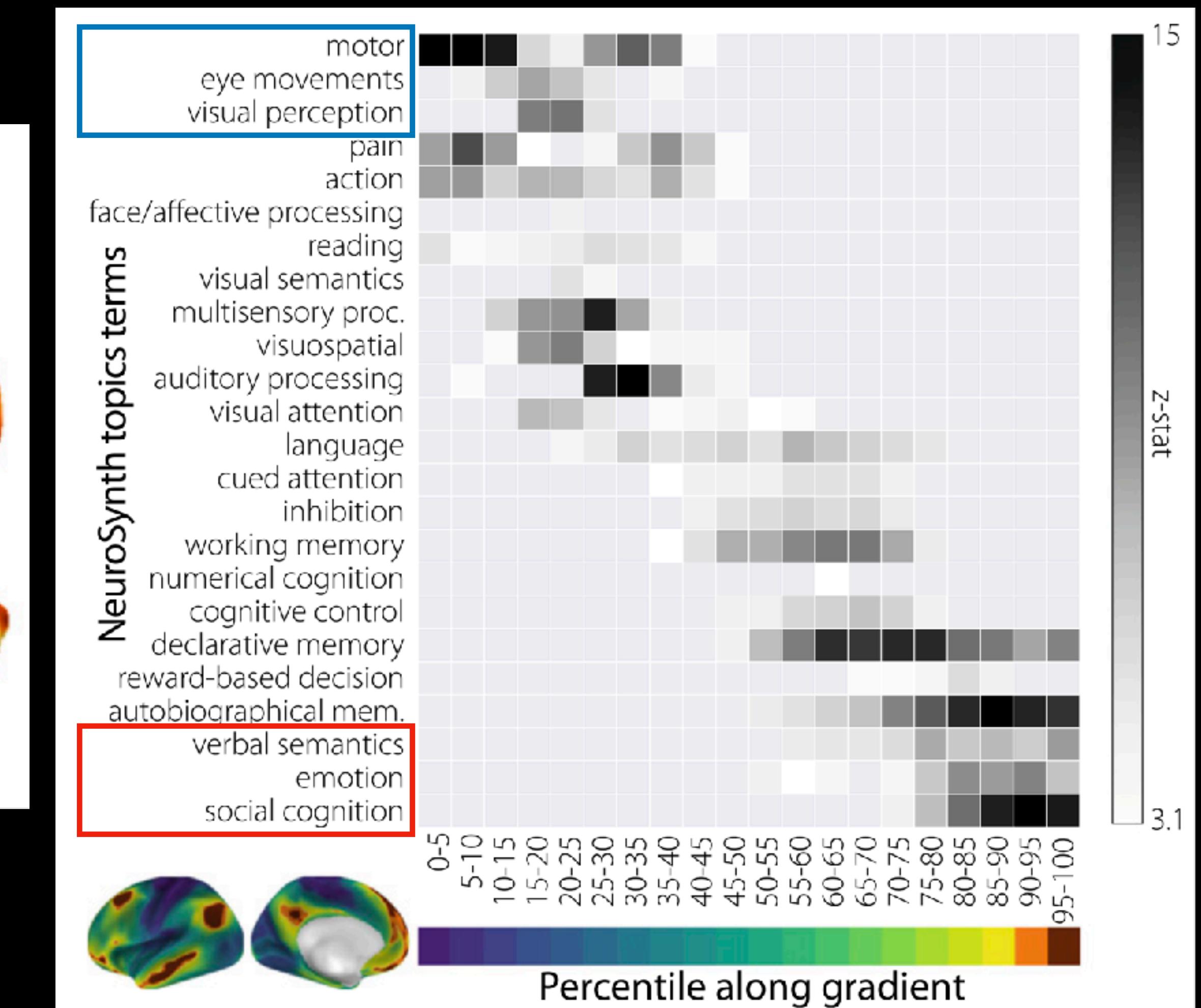


What is already known about the cortical topography of information abstraction?

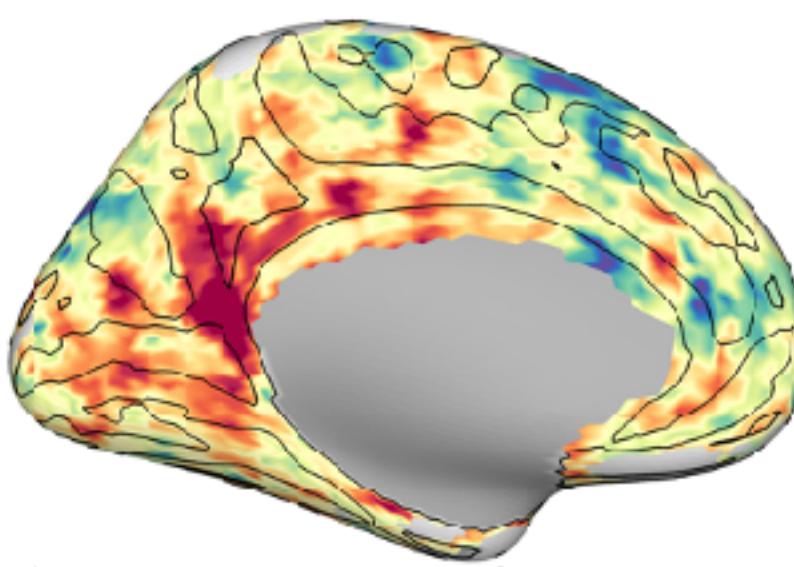
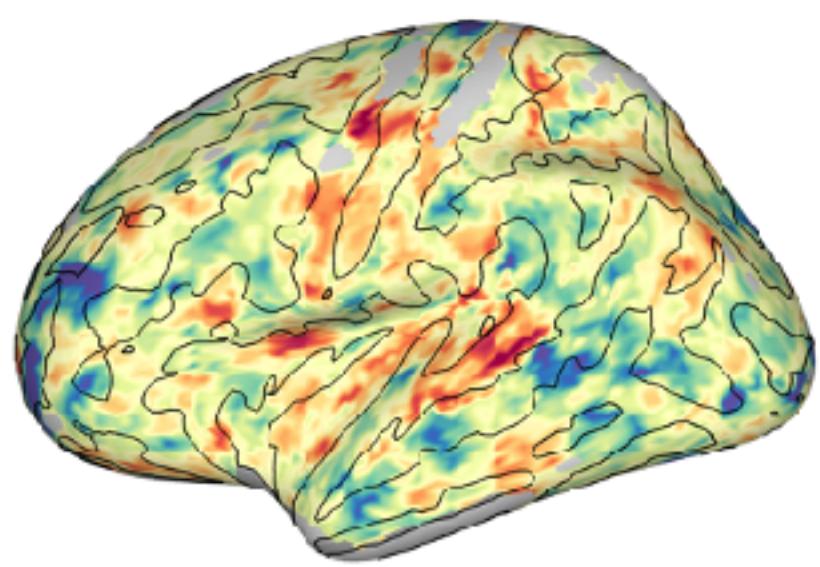
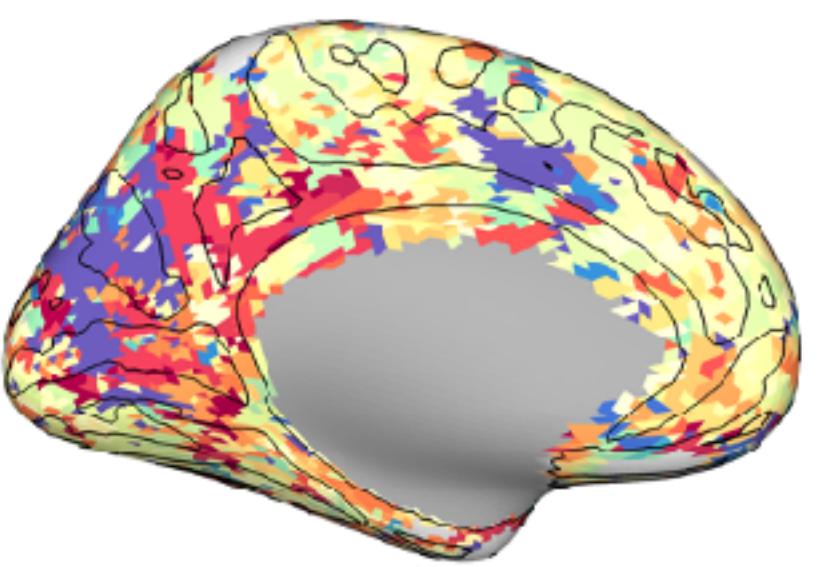
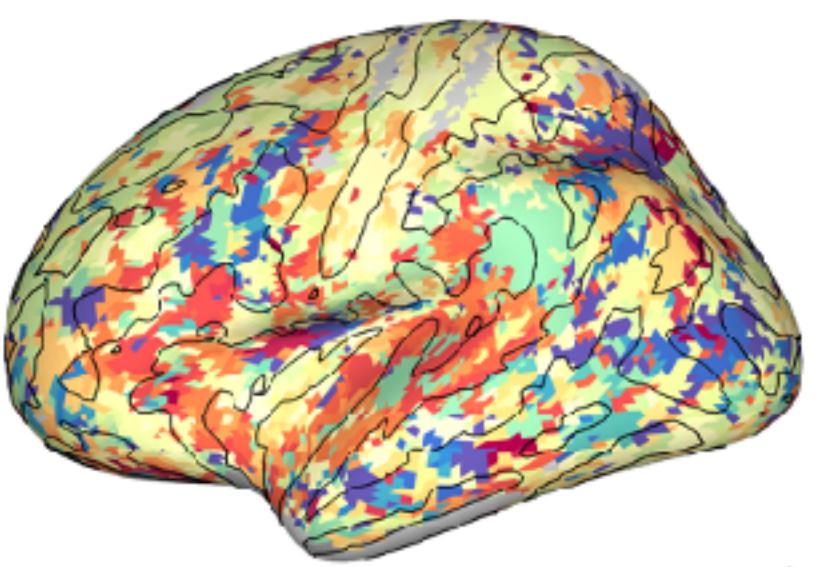
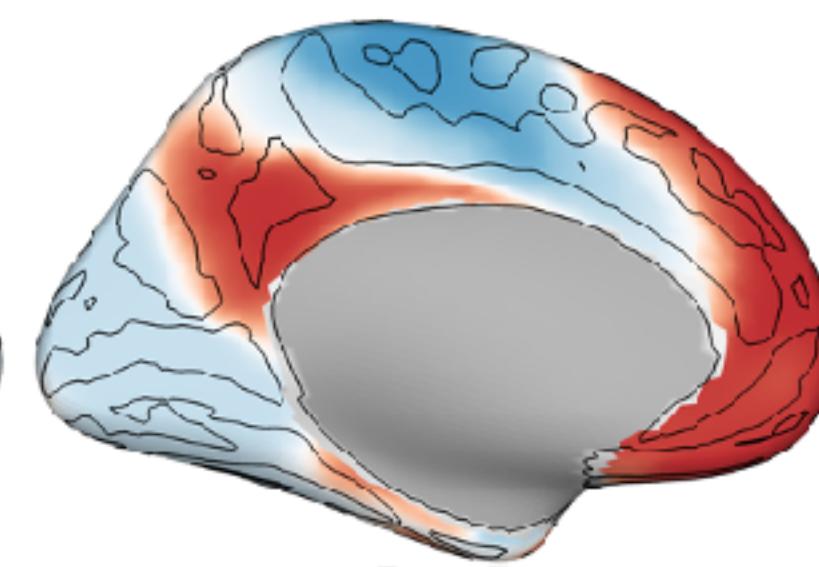
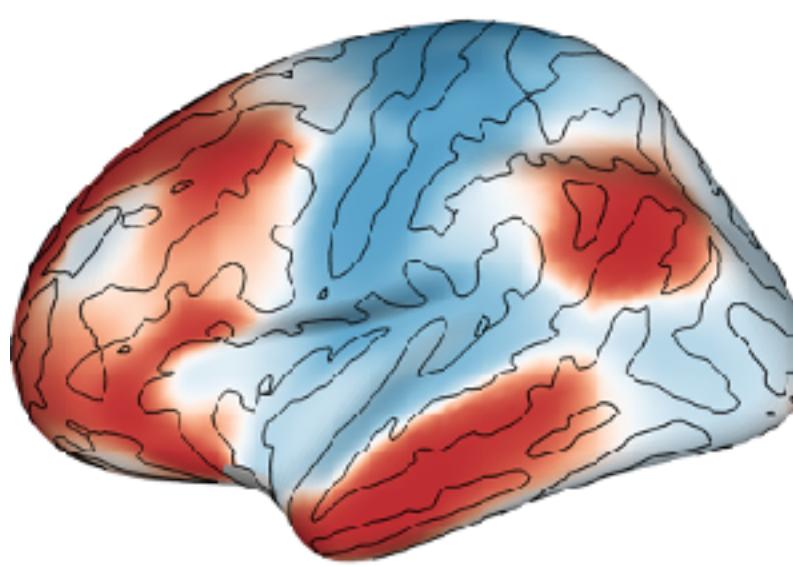
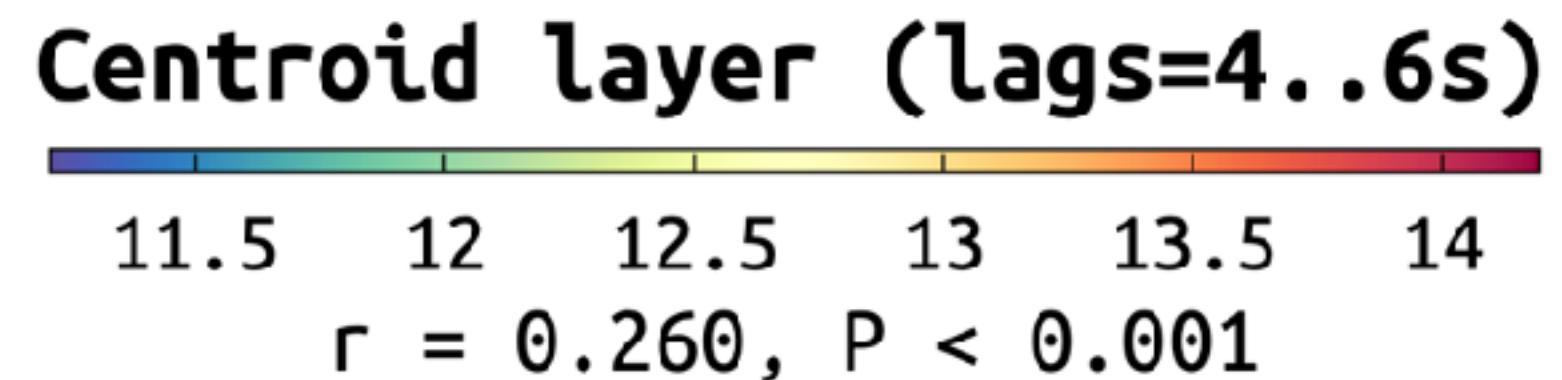
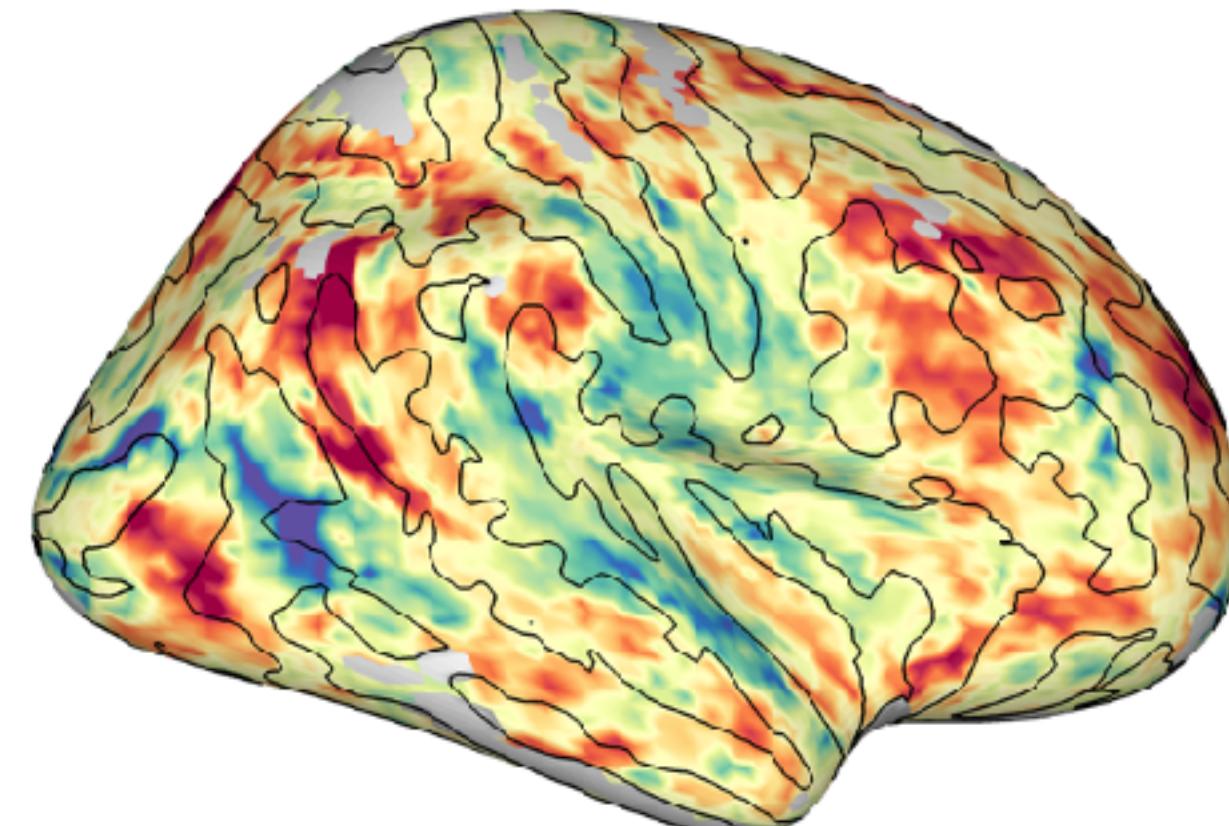
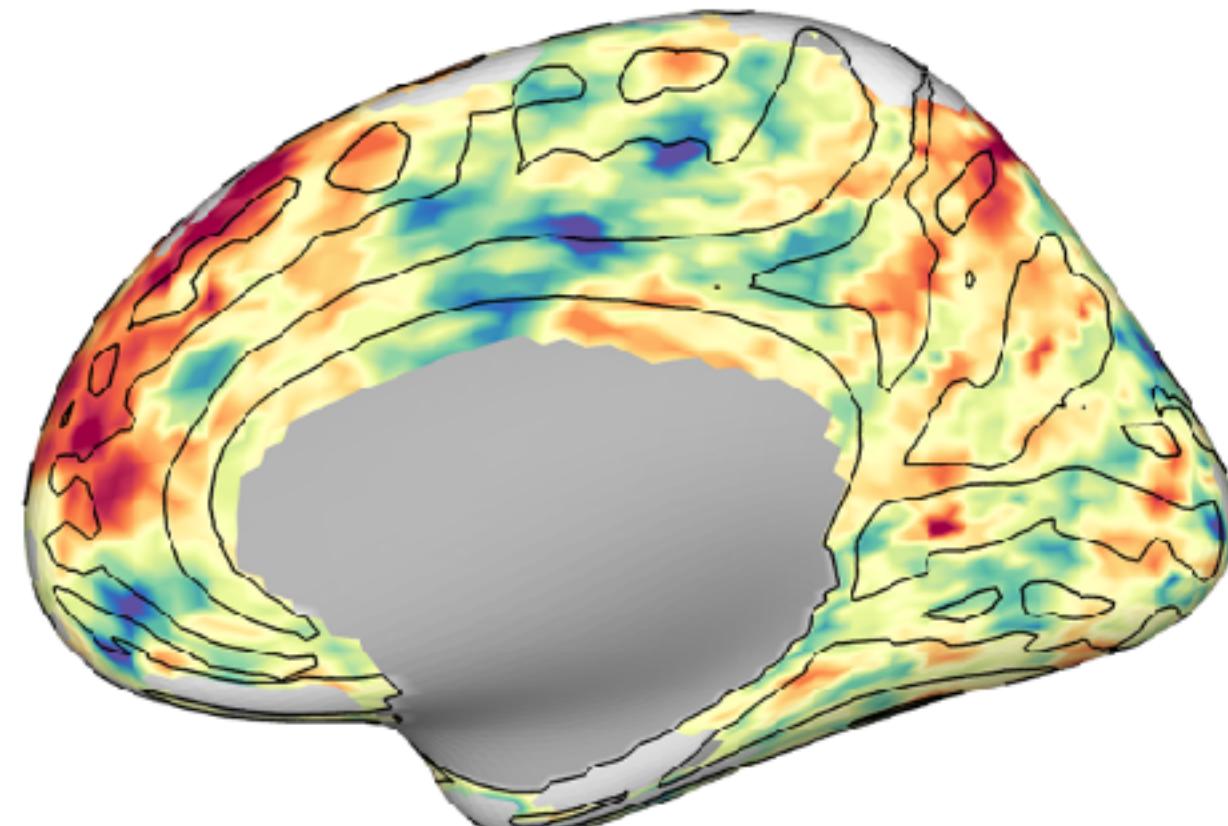
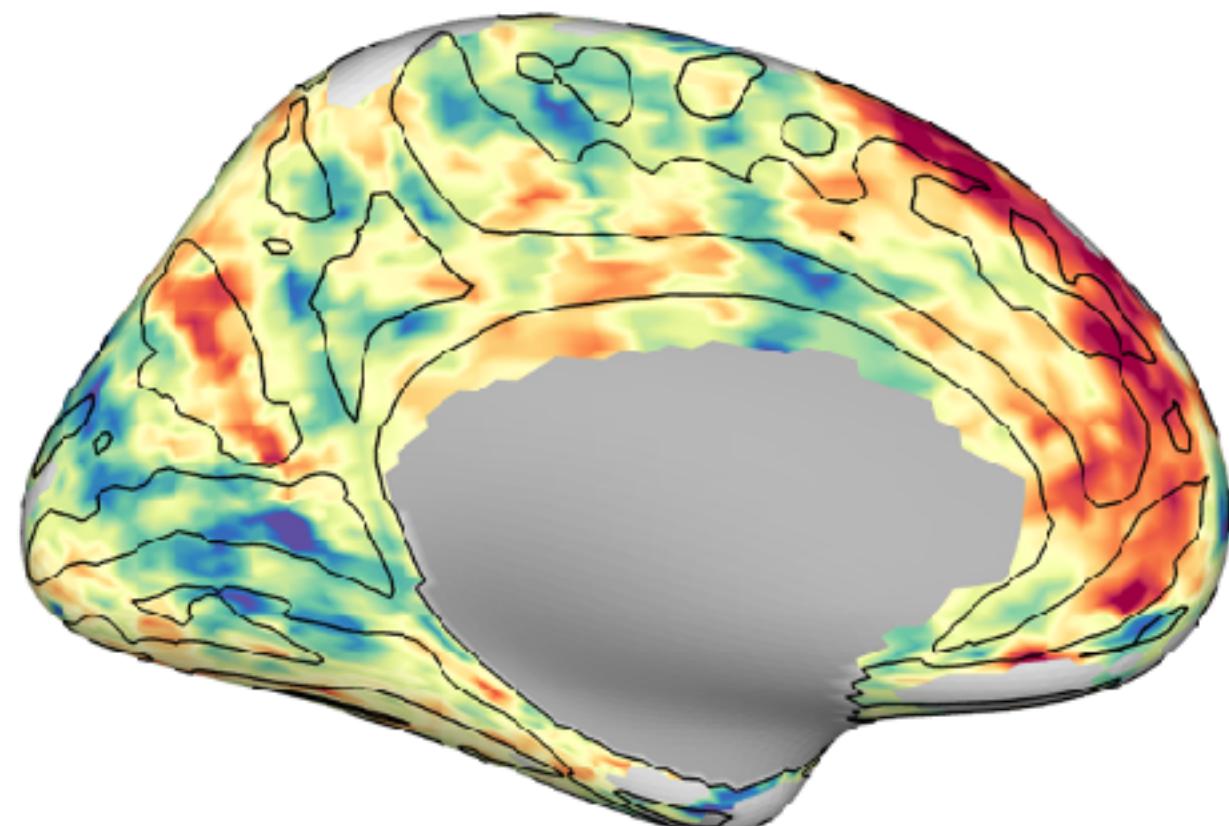
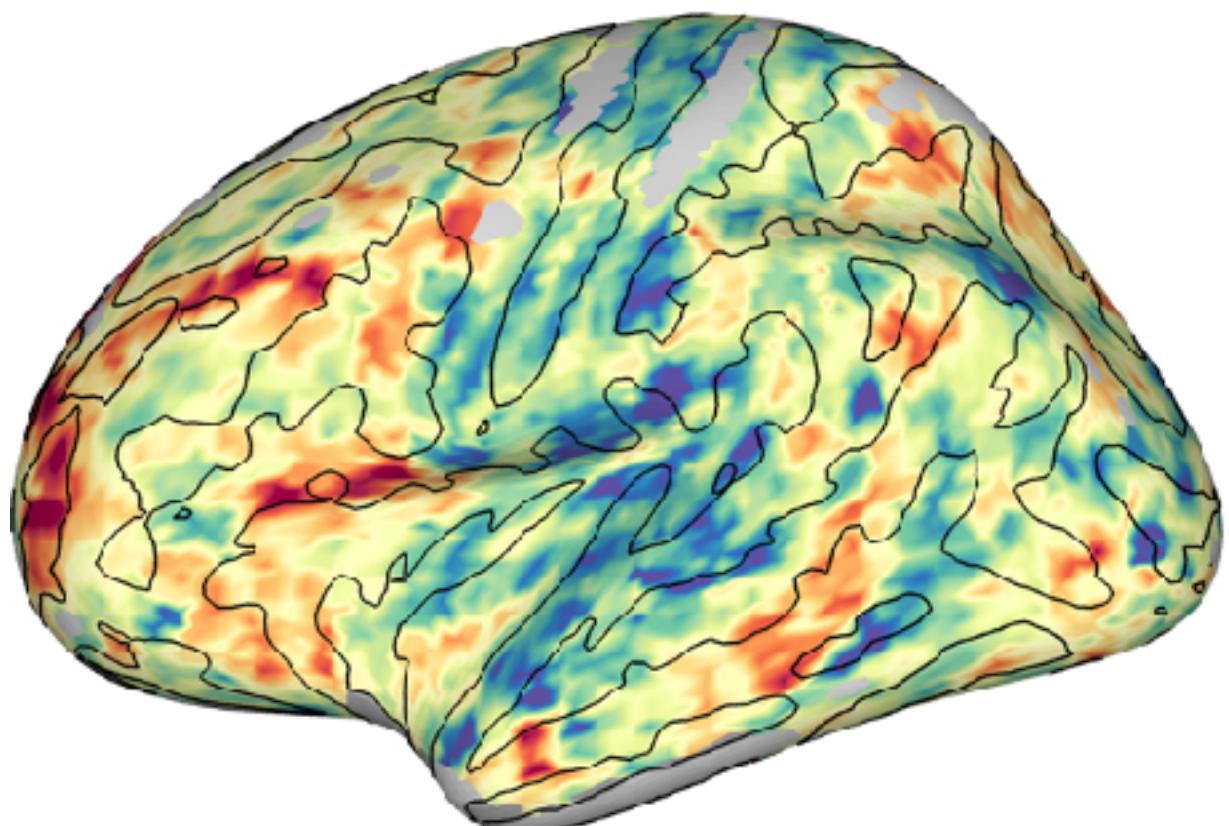
Margulies et al., 2016, PNAS



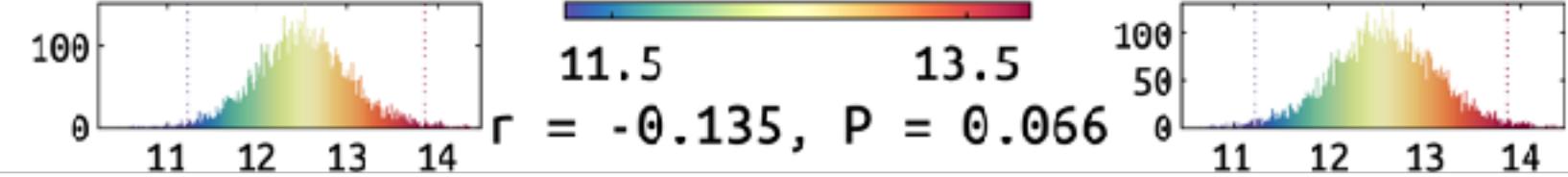
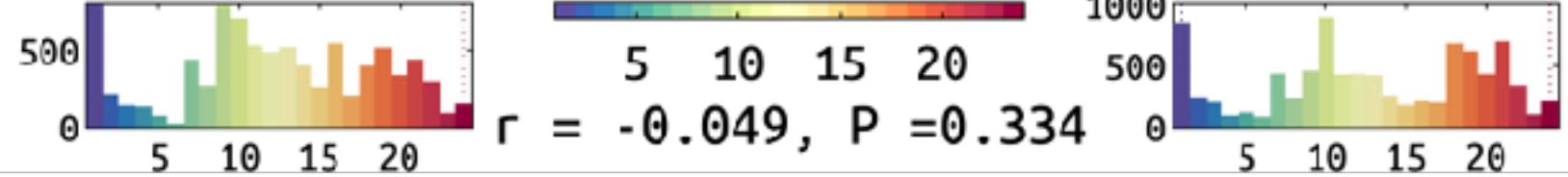
Topography found from resting-state functional connectivity



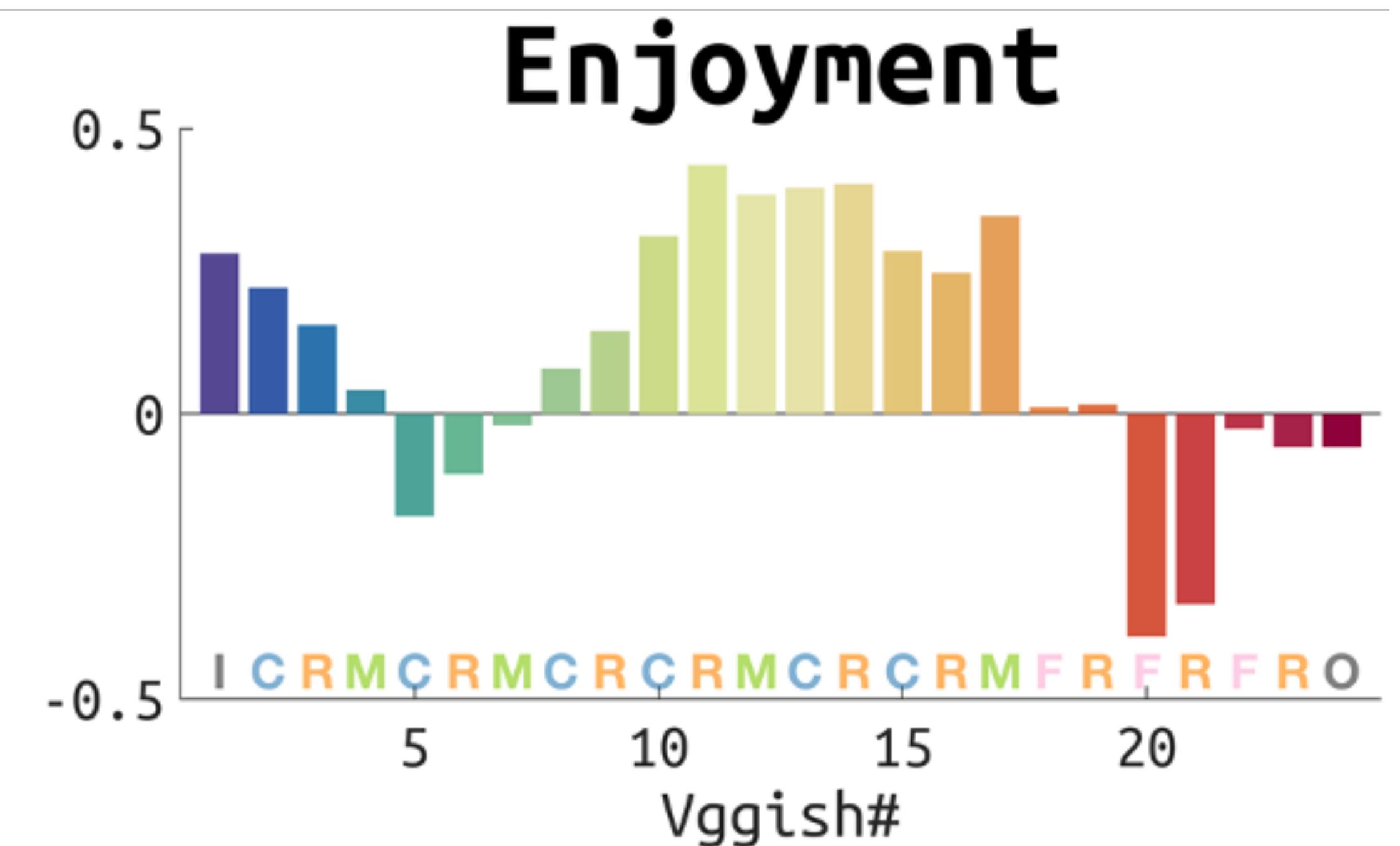
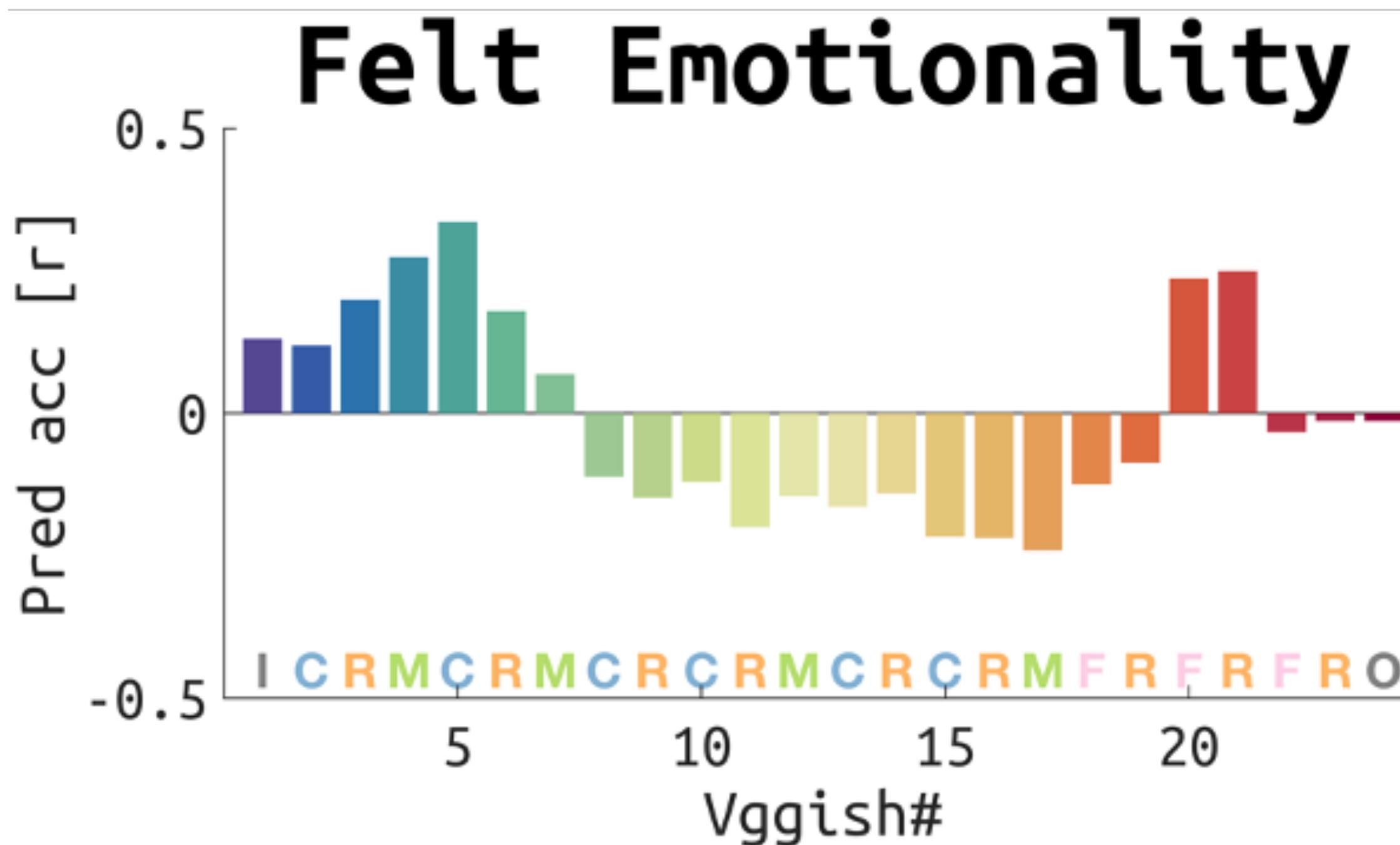
vs. the known topography of abstraction



Functional gradient #1 [au]

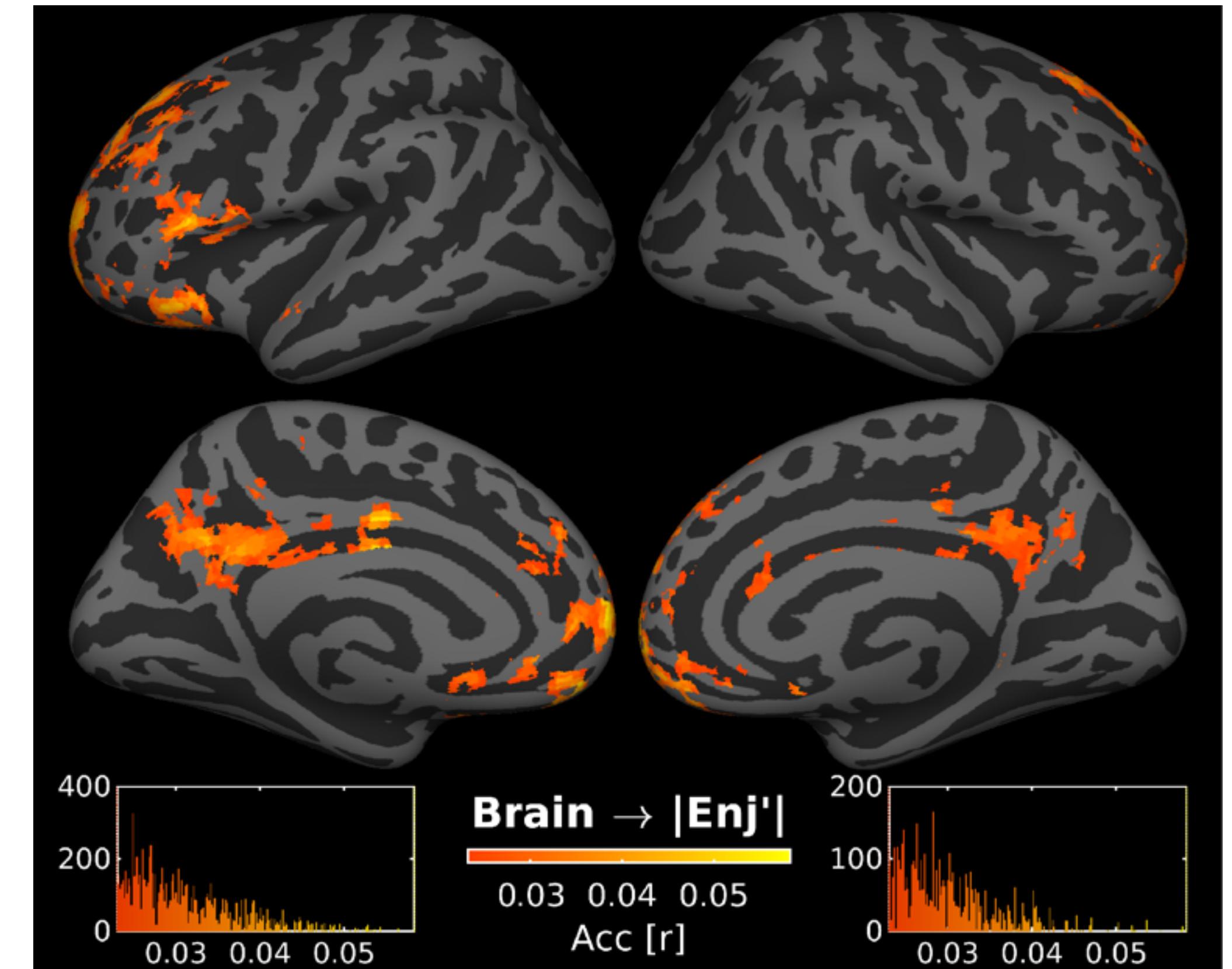
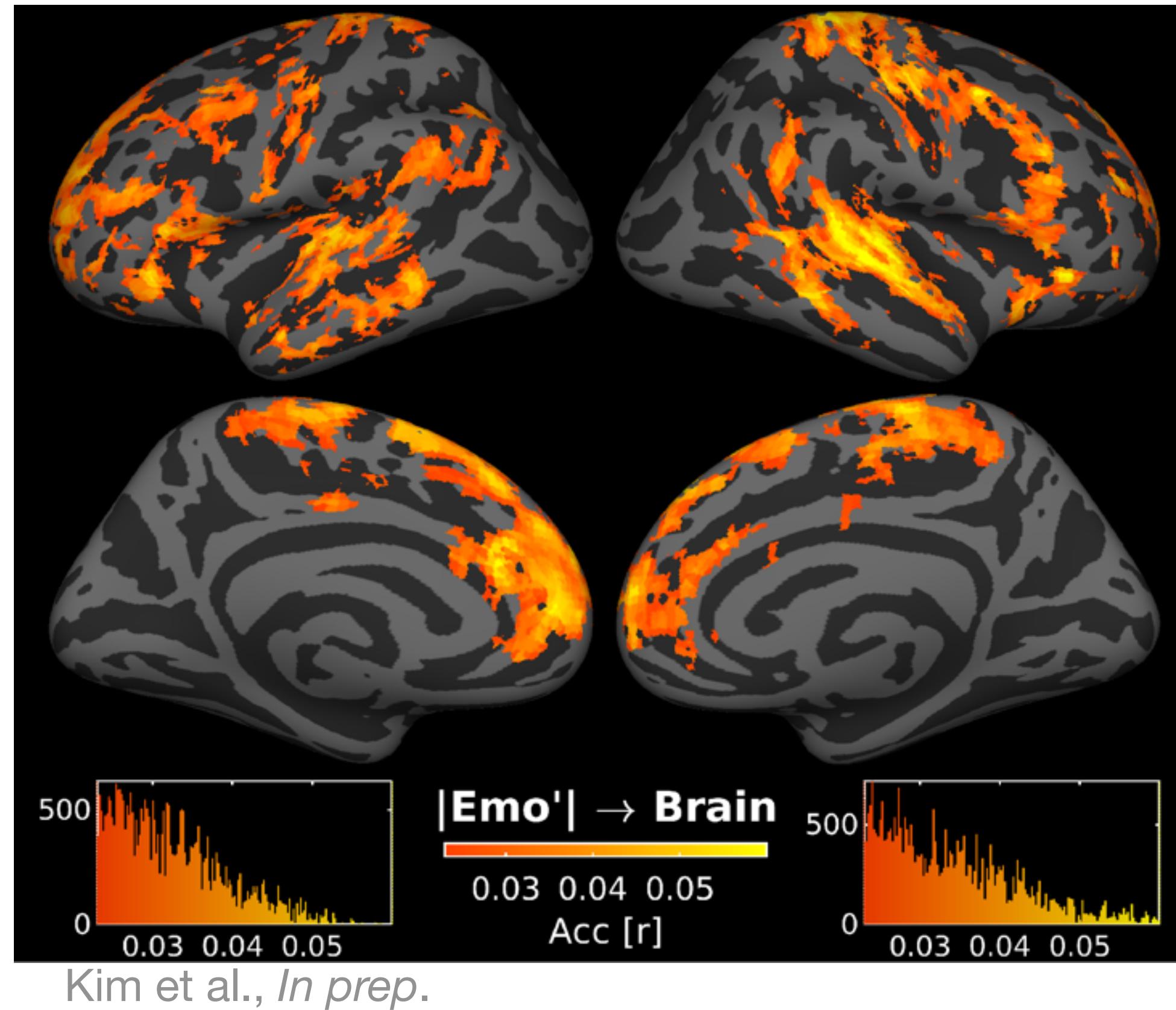


Distinctive patterns of Emotionality and Enjoyment 😊😢 → 🧠



Brain-emotion correlation 😊↔️🧠

fMRI = |Emotionality| + |Enjoyment| + error

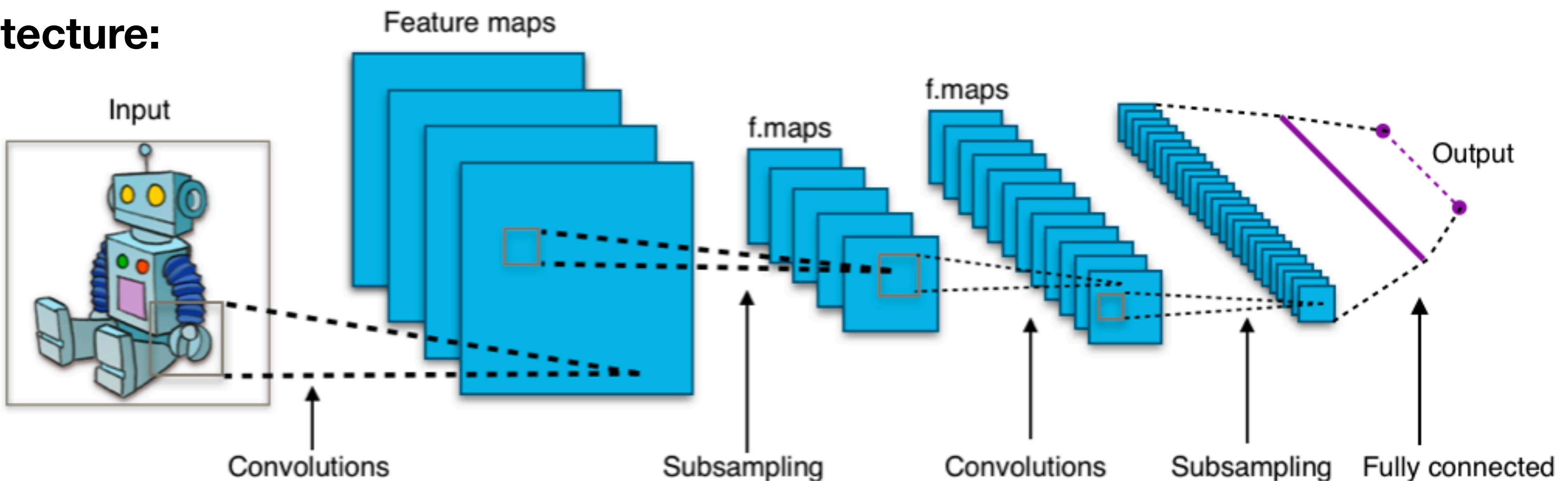




Discussion

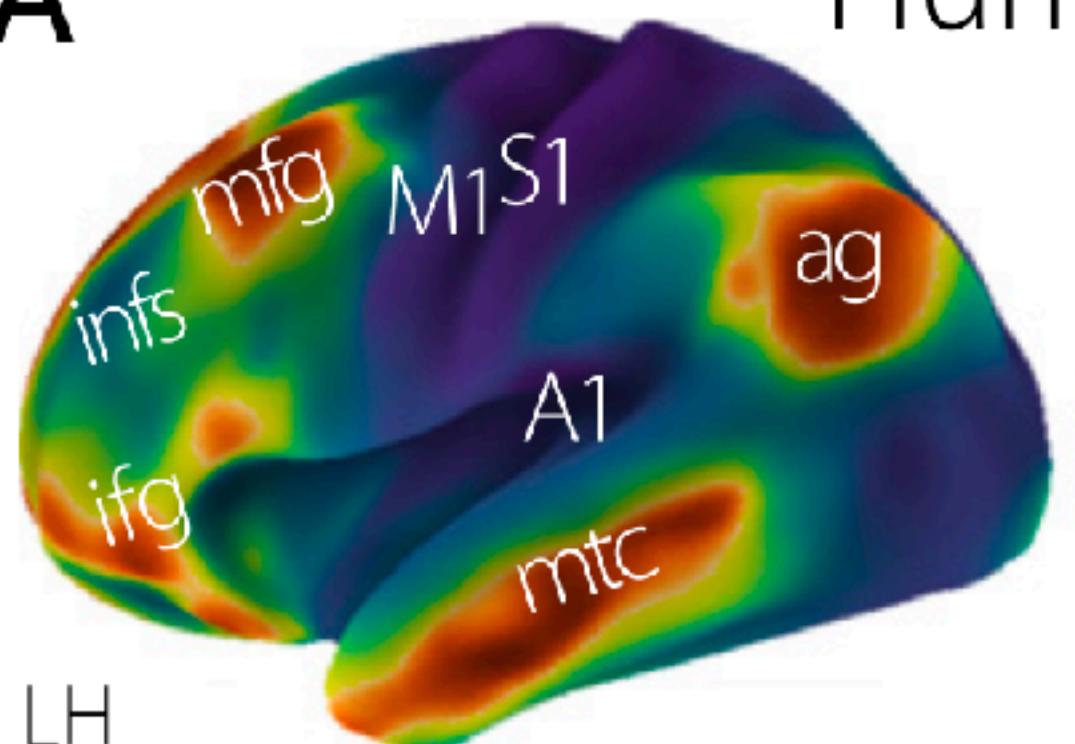
Abstraction of sounds

A CNN architecture:

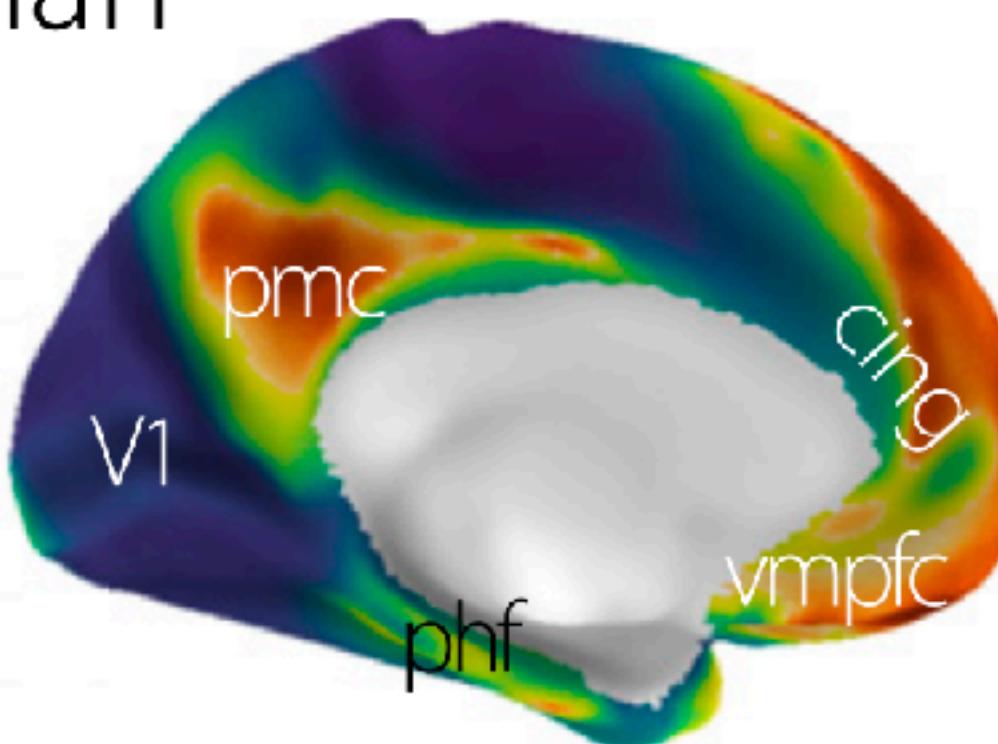


Abstraction in the human cerebral cortex:

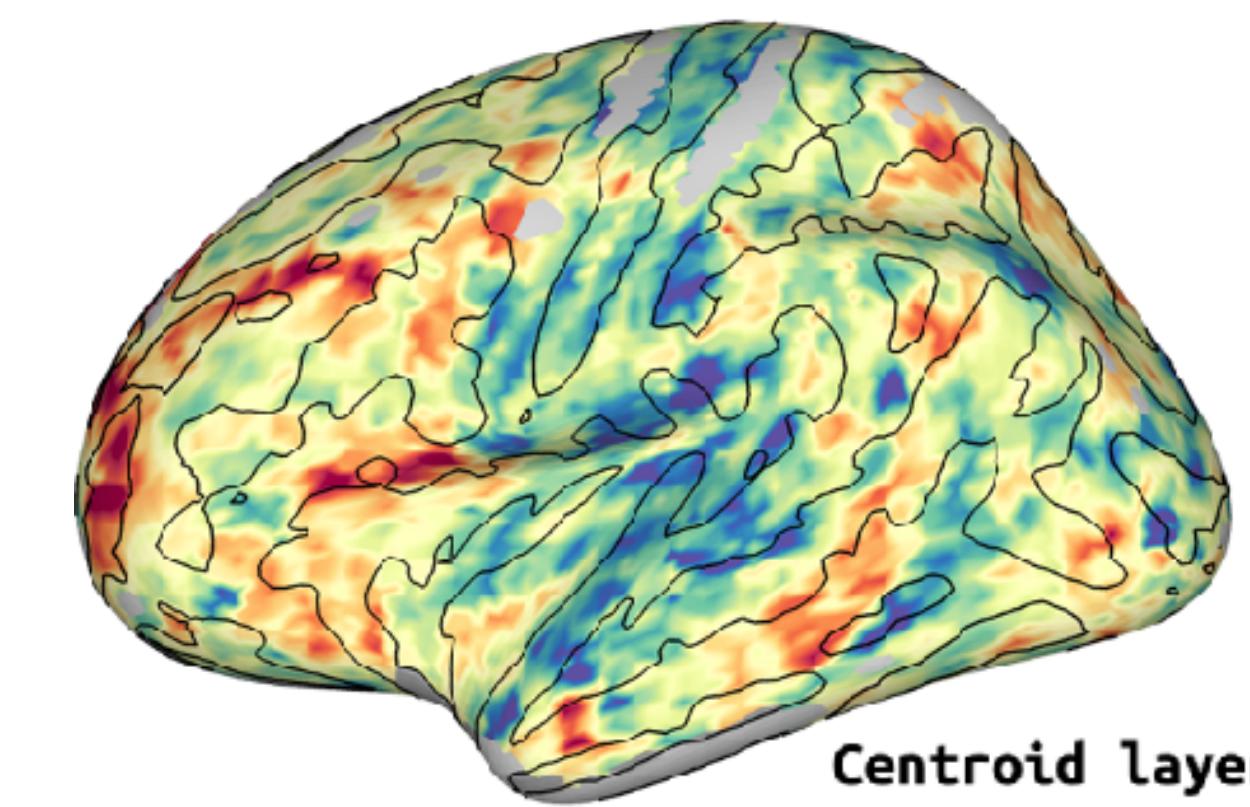
A Human



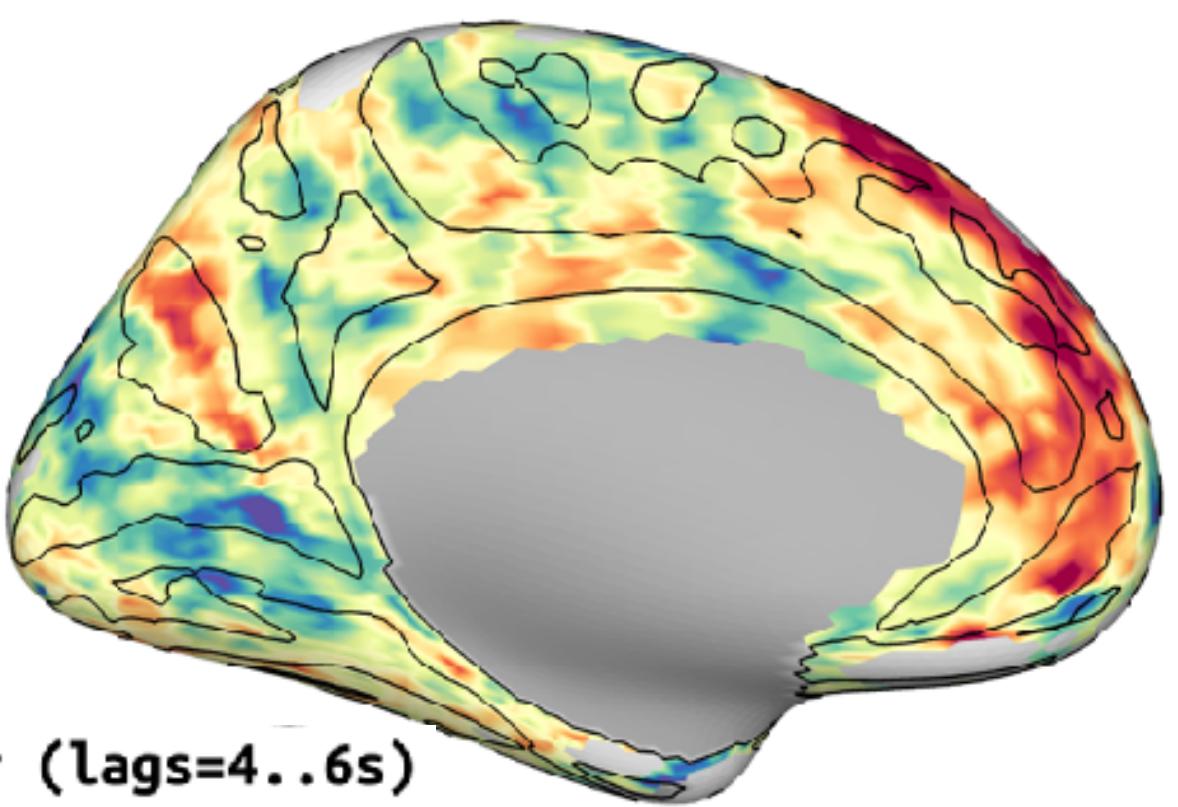
Marguleius et al., 2016, PNAS



Musical abstraction in the human cerebral cortex:

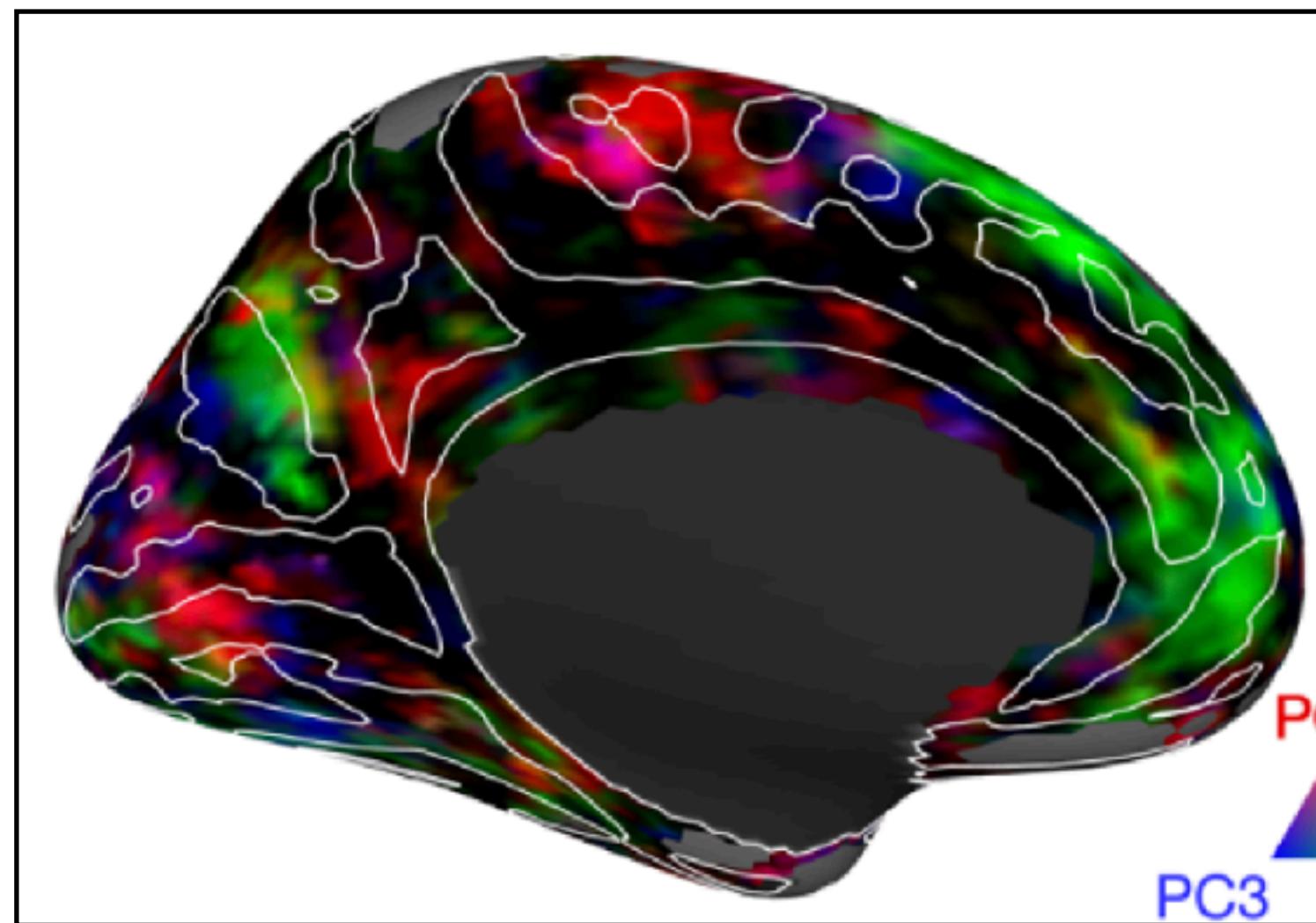


Centroid layer (lags=4..6s)
11.5 12 12.5 13 13.5 14

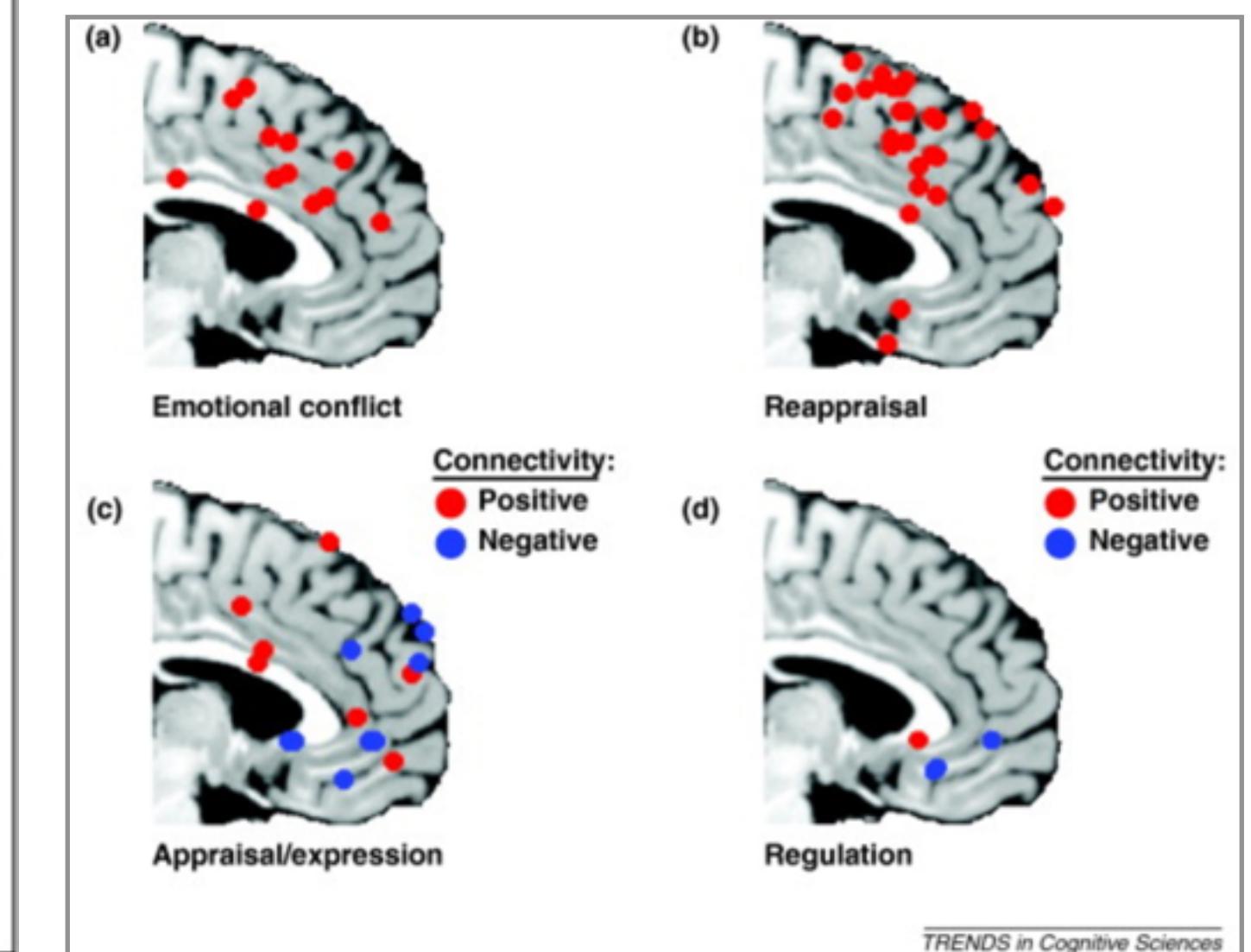
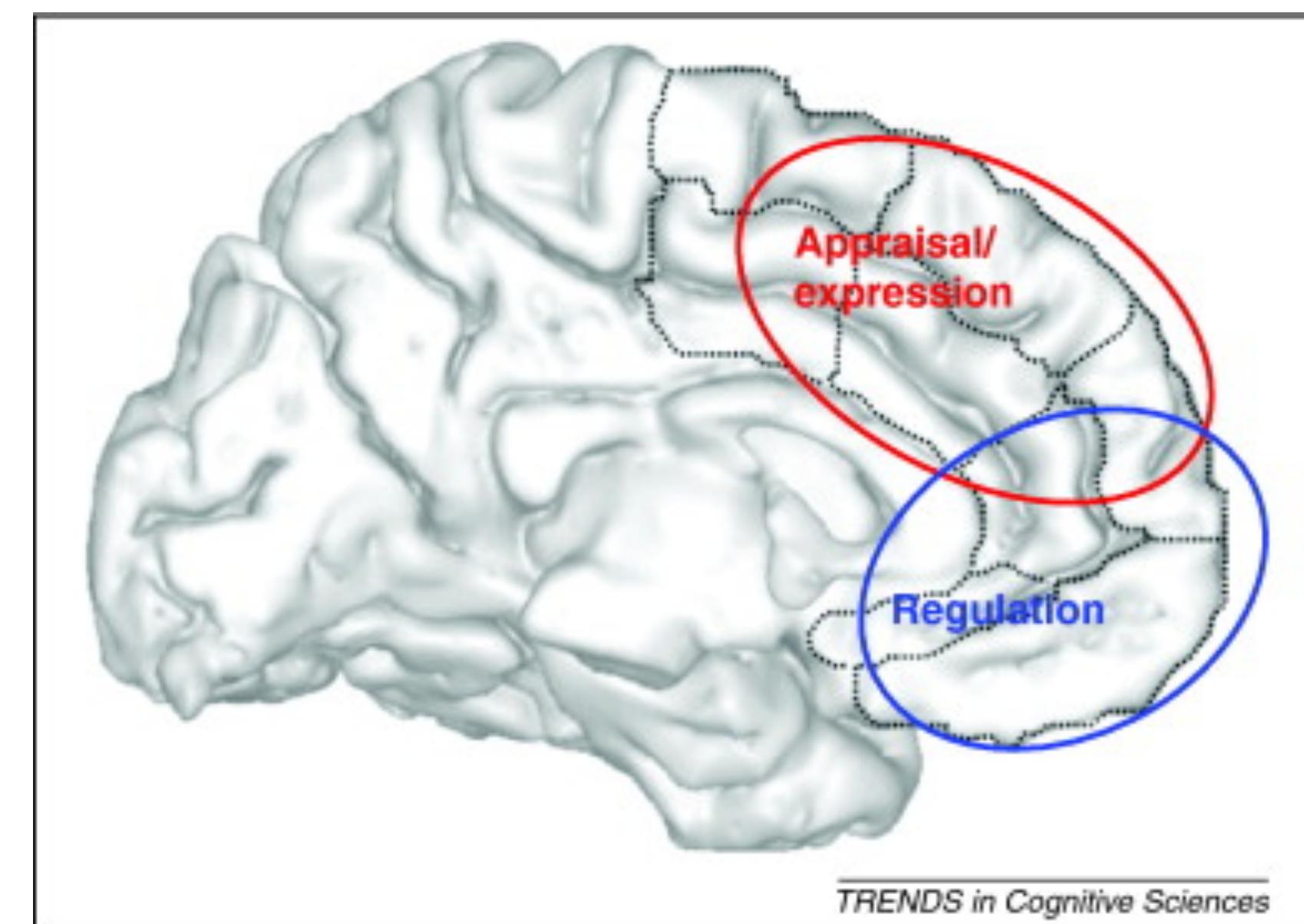


Kim et al., In prep.

mPFC and emotional processing



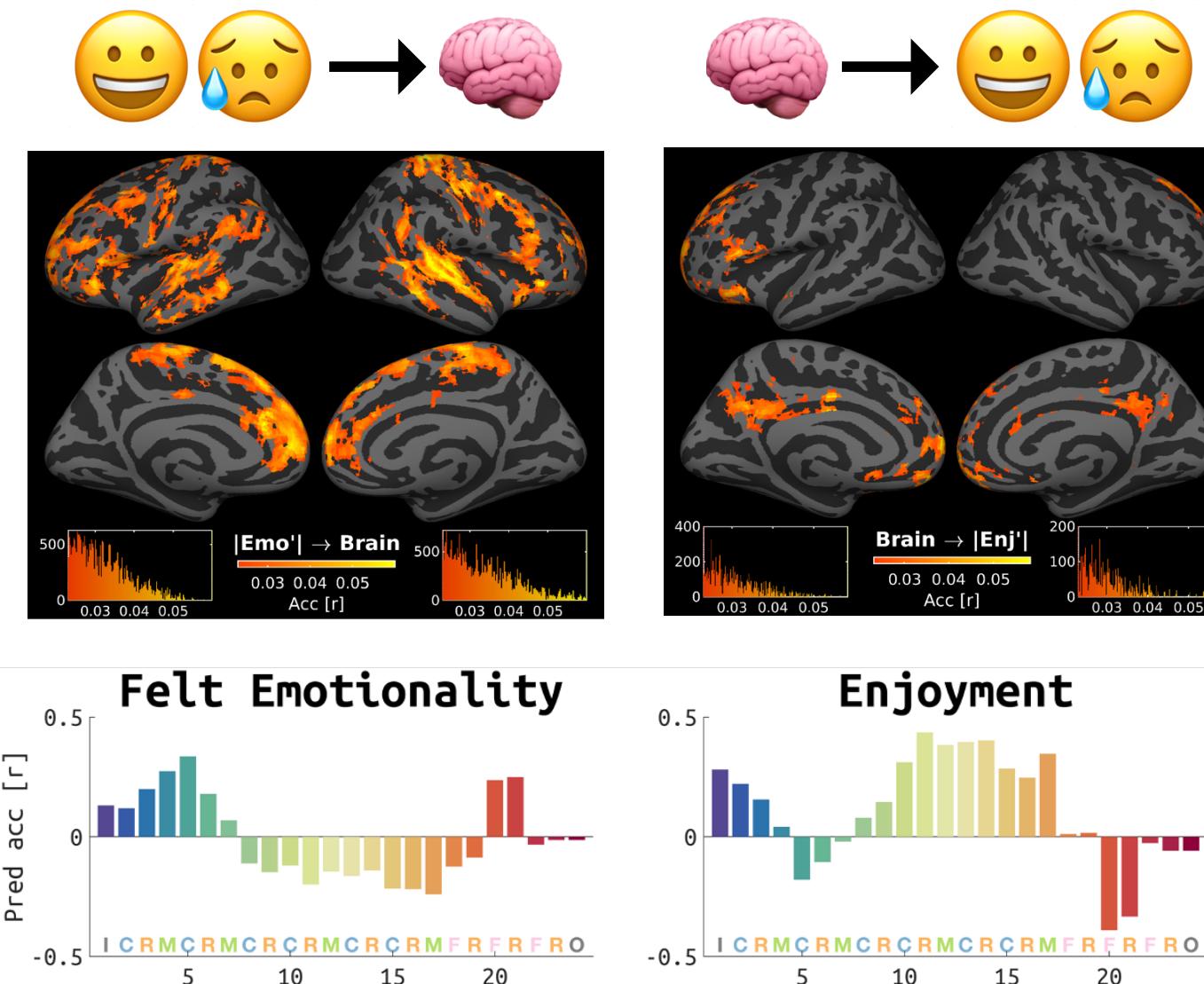
Kim et al., In prep.



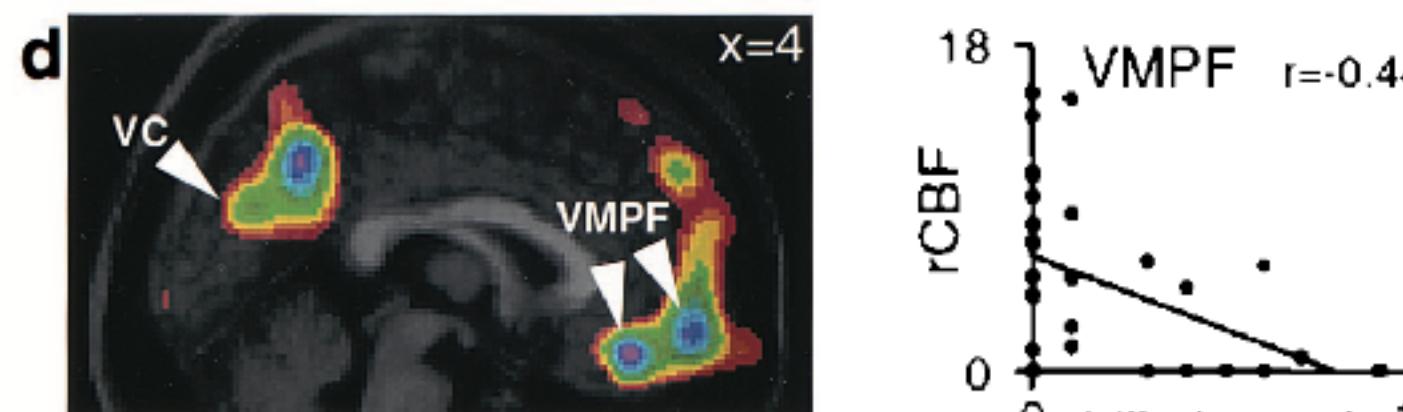
Etkin, Egner, Kalisch, 2010, *Trends in Cognitive Sciences*.

Audio semantic model changes were encoded in the mPFC, which showed a sensitivity to musical structures ("boundaries").

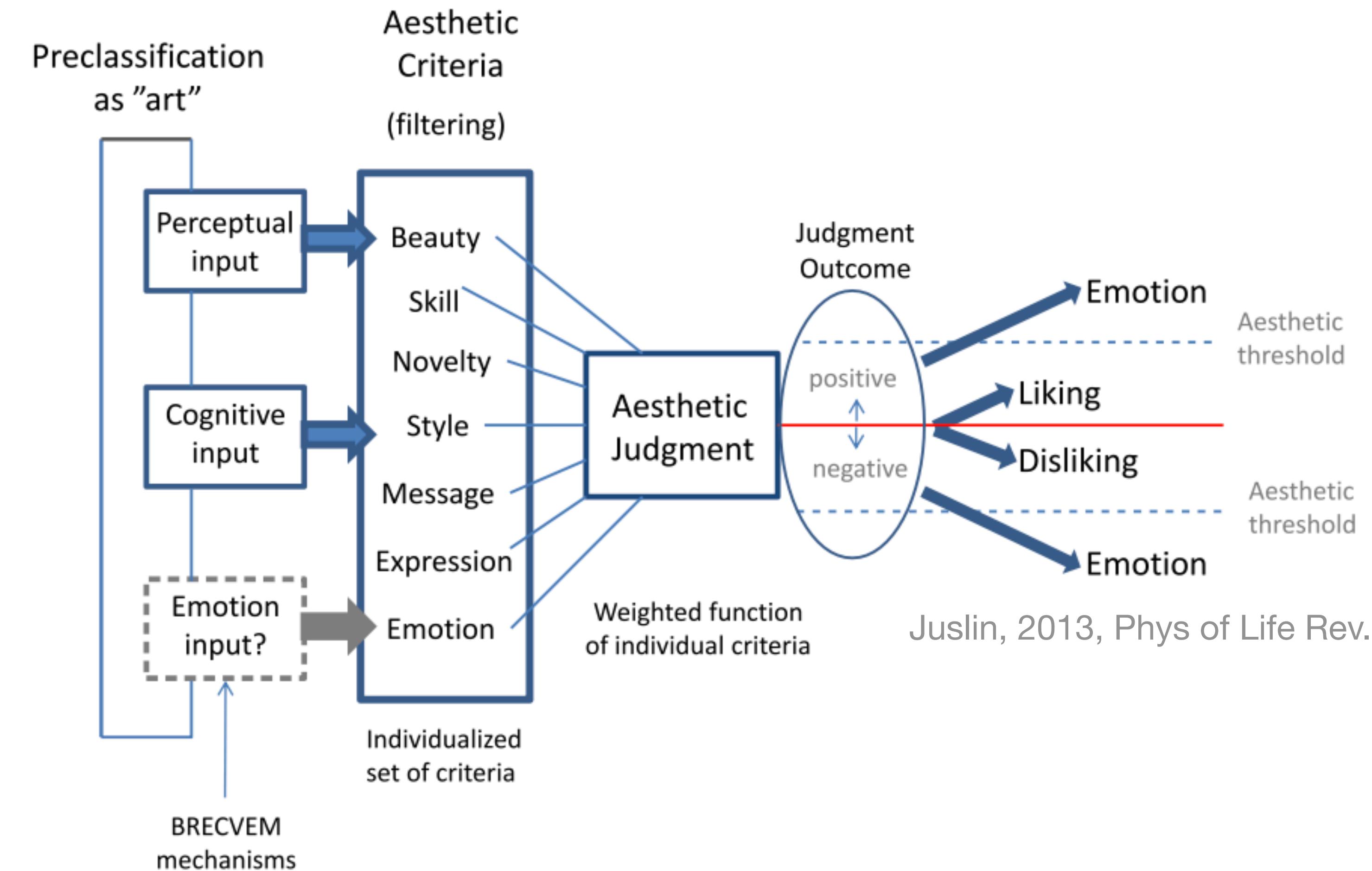
Different encoding of emotionality & enjoyment



Kim et al., In prep.



Blood & Zatorre, 2001, PNAS



Juslin, 2013, Phys of Life Rev.

vmPFC activity was followed by Enjoyment rating changes.

Conclusions

- **Naturalistic stimuli** with computational models allow ecologically valid investigation of evoked emotions.
- CNN embeddings are sensitive to information that is relevant for emotional responses, **beyond low-level audio features**. In particular, the abstraction in CNN shows high similarity to **the known cortical topography of information abstraction**, as well as relevance to behavioral ratings of emotional responses.
- Two continuous ratings (*Emotionality* and *Enjoyment*) were differentially encoded in the brain and predicted by different CNN layers, potentially reflecting **distinct mechanisms of *felt emotions* and *aesthetic judgements***.

Thank you for your attention! (and time for discussion) 😎

<https://seunggookim.github.io/>



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<https://www.aesthetics.mpg.de/en/research/research-group-neurocognition-of-music-and-language.html>