

Harnack-Haus, Berlin | 5 December, 2024



MAX PLANCK INSTITUTE
FOR EMPIRICAL AESTHETICS



Cognition Academies, Max Planck School of Cognition

Music, Brain, and Emotions: a practical workshop

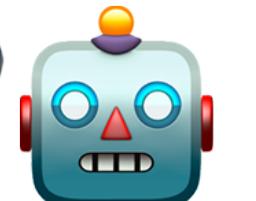
Seung-Goo ("Sng-Gu") Kim

Research Group Neurocognition of Music and Language, Max Planck Institute for
Empirical Aesthetics, Frankfurt am Main



: Who is this?

Seung-Goo Kim

- Originally from Seoul, South Korea, been in Germany for 8+ years without learning German 😜
- First name: Seung-Goo 승구[/swŋgu/](勝九)
 - You can always call me “SG” [Es-Gee] if you feel [w] is too difficult. 😊
- Family name: Kim 김[/gim/](金)
- And very into   



2016, Leipzig



2019, Jeju



Song
Instrument
Order
Pattern

Armani Showers
01: (C) Edge / EMF 1994
000/010
00

Row 00

Channel 01

File
Chord
C. Tempo
C. Speed

armani.s3m (S3M)
none
7D
06

Baseoctave | 3

ESC Main Menu
F10 Quick-Help
CTRL-Q Quit to DOS

F1 . F4 . . . Edit Screen
CTRL-L . . . Load Module
F5/F8 . . . Play / Stop

FreeMem: 281K
FreeEMS: 31280K
FreeGUS: 1018K

Pattern Editor (F2)

01: L1

02: R1

03: L2

04: R2

05: L3

00	E-3	01	01	S85
01	A-3		01	GAF
02	B-3		02	G00
03	E-3		02	G00
04	B-3		03	G00
05	C-4		03	G00
06	E-3		04	G00
07	D-4		05	G00
08	E-3		05	G00
09	A-3		06	G00
10	B-3		06	G00
11	E-3		07	G00
12	B-3		07	G00
13	C-4		08	G00
14	E-3		09	G00
15	E-4		09	G00
16	E-3		10	G00
17	A-3		10	G00
18	B-3		11	G00
19	E-3		11	G00
20	B-3		12	G00
21	C-4		13	G00
22	E-3		13	G00
23	D-4		14	G00
24	E-3		14	G00
25	A-3		15	G00
26	B-3		15	G00
27	E-3		16	G00
28	B-3		17	G00

E-4 01 00 S8F

01	.	00	.	00
02	.	.	.	00
03	.	.	.	00
04	.	.	.	00

00	.	00	.	00
01	.	01	.	00
02	.	02	.	00
03	.	03	.	00

00	.	00	.	00
01	.	01	.	00
02	.	02	.	00
03	.	03	.	00

$$F = \frac{ma}{\sqrt{1-\mu^2/c^2}} + \frac{m \cdot (\mu c) \gamma c^2}{(-\mu^2/c^2)^2}$$

$$\lim_{\Delta y \rightarrow 0}$$

$$f(x_0, y_0 + \Delta y) - f(x_0, y_0)$$

$$\Delta y$$

$$2+2=4 \quad \Delta = \sqrt{P(P-a)(P-b)(P-c)} = P \cdot r \cdot A_2O$$

$$AB = \sqrt{(x_2-x_1)^2 + (y_2-y_1)^2}$$

$$ax + bx + cx = 0$$

$$E = mc^2$$

$$a^2 - b^2 = (a-b)(a+b)$$

$$h = \sqrt{a^2 - b^2} = \sqrt{a^2 - b^2} = \sqrt{a^2 - b^2} = \sqrt{a^2 - b^2}$$

$$f(x) = a(x-x_1)(x-x_2)$$

$$C(x) = a(x-x_1)(x-x_2) \frac{b^m}{a^m} = \frac{b^m}{a^m} \sqrt{a^2 - b^2} = \sqrt{a^2 - b^2} = \sqrt{a^2 - b^2}$$

$$Cl \text{ (benzene ring)} + H_2O \rightarrow HCl + OH^-$$

$$\cos \alpha + \cos \beta = 2 \cos \frac{\alpha + \beta}{2} \cos \frac{\alpha - \beta}{2}$$

$$Z = \frac{1}{\sqrt{2\pi}} e^{-\frac{r^2}{2}}$$

$$r = \sqrt{x^2 + y^2}$$

$$\sin \alpha + \sin \beta = 2 \sin \frac{\alpha + \beta}{2} \cos \frac{\alpha - \beta}{2}$$

$$b = \frac{\log c}{\log a}$$

$$\log a = \frac{\log c}{\log b}$$

$$dA = \frac{1}{\sqrt{2\pi}} e^{-\frac{B^2}{2}}$$

$$B = \sqrt{B^2 - dA^2}$$

$$Z = \frac{1}{\sqrt{2\pi}} \theta \cdot \frac{d^2}{2}$$

$$2+2=4 \quad E = mc^2$$

$$\int_0^\infty \operatorname{erf}(\sqrt{x}) dx = \sqrt{2}$$

$$2n\alpha = \frac{g_2 - g_1}{2}$$

$$\int_0^\infty \operatorname{erf}(\sqrt{x}) dx = \sqrt{2}$$

Education & Training (1)

- 1996–1999, Musical composition at *Sunwha* Arts High School, South Korea
- 2000–2004, dual BAs in Economics & Psychology, *Yonsei* University, South Korea
- 2008–2010, MS in Cognitive Sciences, Seoul National University, South Korea (MEG for music)
- 2010–2012, Graduate researcher, Seoul National University, South Korea (sMRI for stat)
- 2012–2017, Ph.D. (Dr. rer. nat.) in Psychology, IMPRS for Neuro-Comm, MPI for CBS, Leipzig (s/fMRI for music)

From Wilhelm Wundt's the first ever experimental psychology lab!



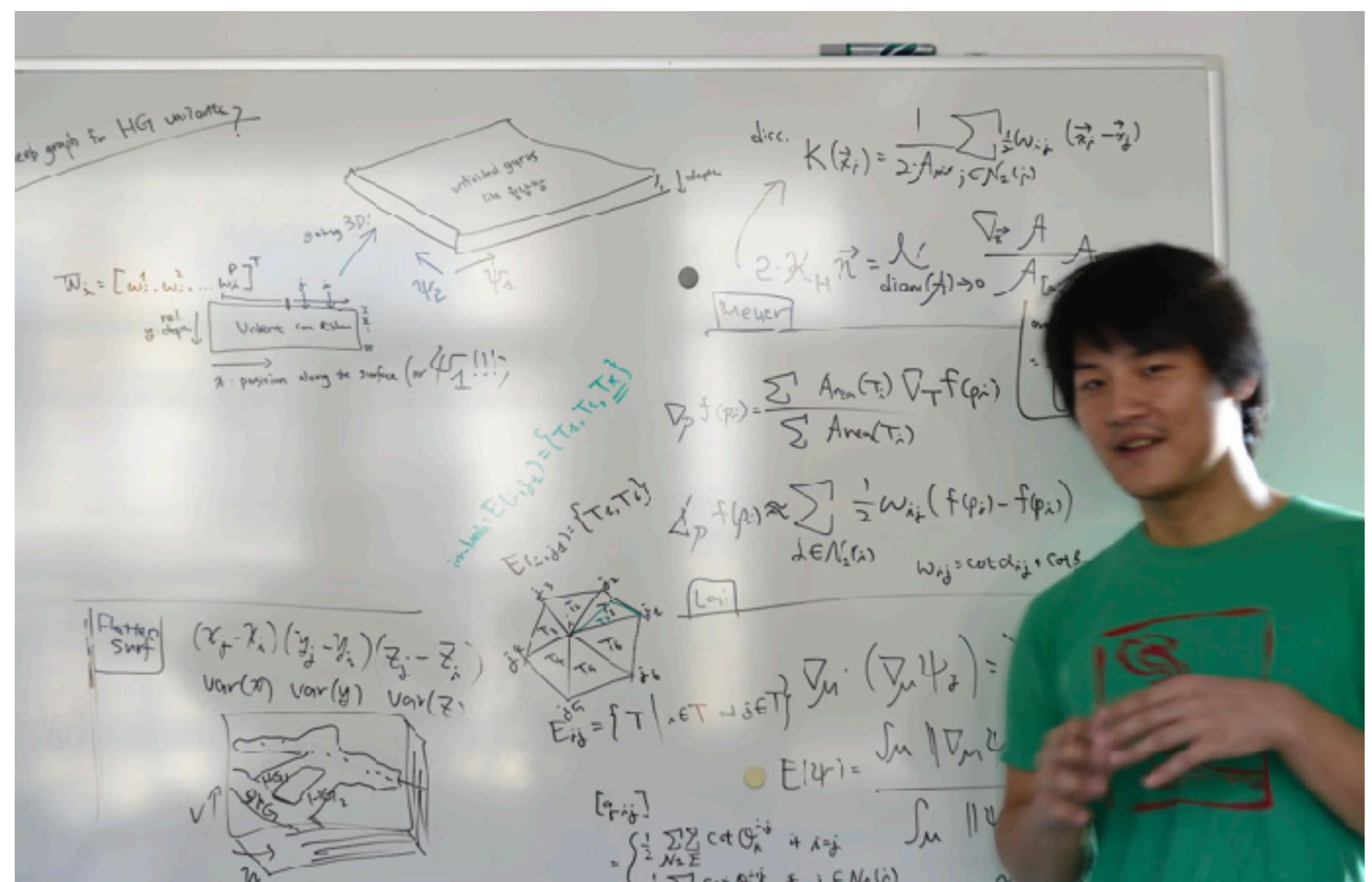
2017, Leipzig



2017, Leipzig

Education & Training (2)

- 2017–2018, Postdoc, Cambridge University, UK (s/fMRI for psychiatry; PI: Dr. Valerie Voon)
- 2018–2021, Postdoc, Duke University, NC, USA (M/EEG, fMRI for speech; PI: Dr. Toibas Overath)
- 2021–, Research Scientist, Max Planck Institute for Empirical Aesthetics, Frankfurt am Main (fMRI for music; PI: Dr. Daniela Sammler)



2013, Leipzig



2018, Durham, NC

MAX PLANCK INSTITUTE FOR EMPIRICAL AESTHETICS







(C) Max Planck Institute for Empirical Aesthetics





**What do we do this
morning?**

Session plan

Music, Brain, and Emotions

From	To	Contents	Dur
09:30	10:15	Introduction: <u>Music, Brain, Emotions</u>	45m
10:15	10:45	Live Experiment: <u>Music & Emotions</u>	30m
10:45	11:00	Coffee break	15m
11:00	11:20	Theories: <u>How to analyze naturalistic experiments</u>	20m
11:20	11:50	Hands-on: <u>Analysis</u>	30m
11:50	12:00	Summary & discussion	10m
		SUM	2h 15m

What's[re] your language[s]?

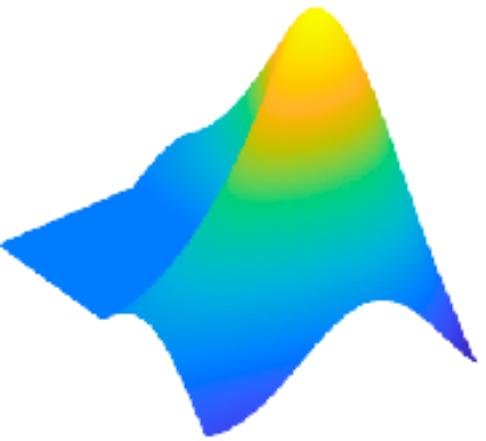
scripting languages (Python, R, Matlab, Julia, Bash, Perl, Lisp)?

compiling languages (Java, C, C++, Rust)?

markup languages (md, tex, html, javascript)?

Excuse my language!

A few words on MATLAB and Open Science



- Why not Python? Have you heard about the Anaconda's Pricing Policy for Max Planck Institutes?
- Using MATLAB does not necessarily harm Open Science (Yes, it may grow dependency; but everyone needs to be **polylingual** anyways!)
- There are a couple of cloud platforms where you can run MATLAB without purchasing the license (e.g. code ocean, MATLAB online, ...)
- Because of the liability, MATLAB is more rigorously maintained than many community codebases.
- Still many people understand MATLAB.
- (Real reason: Sorry, I have all my toys  in MATLAB until I rewrite them in Julia)



Music, Brain, and Emotions

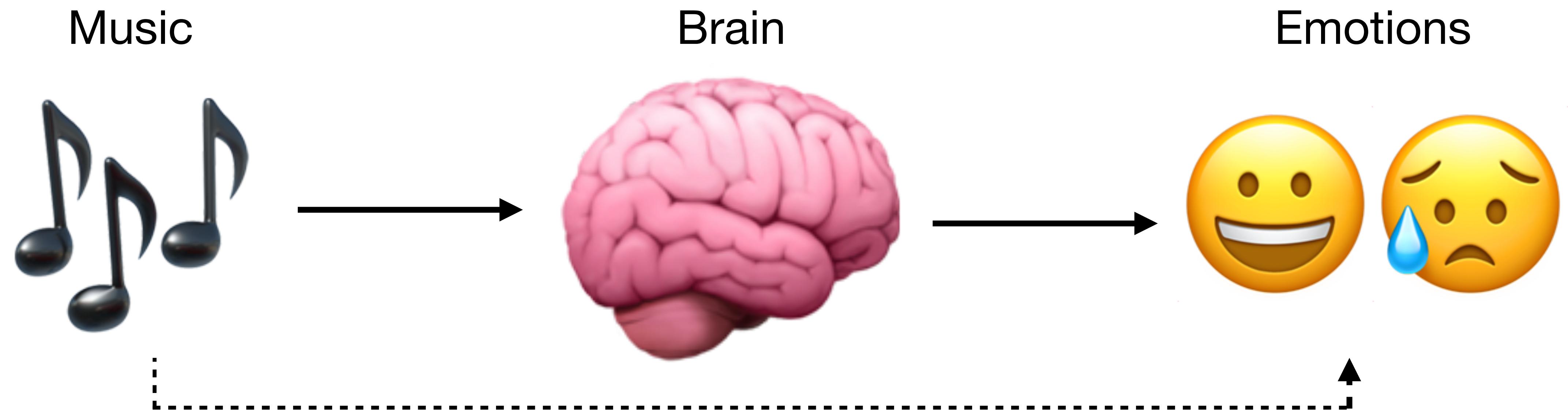
Introduction

SOUND DEMO: "Act I – Prelude", Tristan und Isolde, R. Wagner, Gewandhausorchester, Andris Nelson



How does music evoke emotions *via* the brain?

A very simplistic view

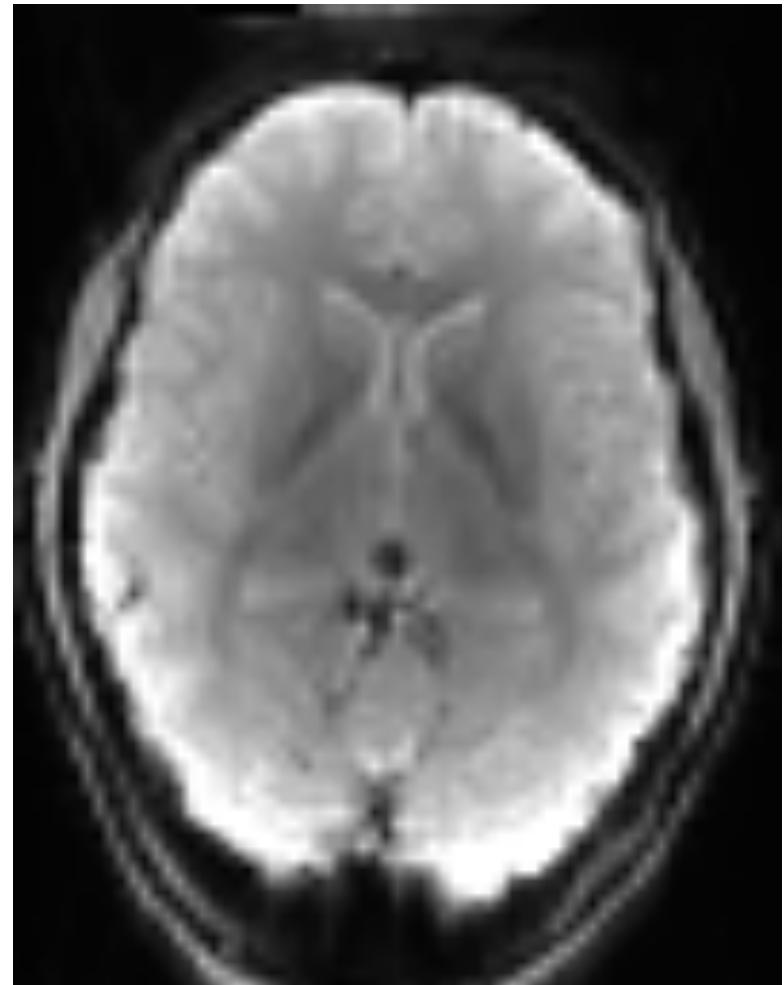


How does the brain transform the musical information?

Functional magnetic resonance imaging (fMRI)

Still, the state-of-the-art of non-invasive neuroimaging

T2* EPI image



- **Neurovascular coupling** {Somehow: probably by glial cells?}: neural activity → blood flow with oxygenated hemoglobin
- Oxygenation makes blood less paramagnetic, increasing the EPI image intensity ("**Blood-Oxygenation-Level-Dependent [BOLD] effect**" [1])
- Neural activity → BOLD response has been experimentally confirmed in animals (electrode-implanted monkeys [2]; optogenetic rodents [3]).
- It's **loud** (90+ dB; ANC: ~-30 dB), noisy (SNR = -2 dB), and expensive (€2M + a team led by an MR physicist). But it's still the SOTA (1 sec per volume & 3 x 3 x 3 mm³ voxel).

[1] Ogawa et al., 1990, *PNAS*. [2] Logothetis et al., 2001, *Nature*. [3] Lee et al., 2010, *Nature*.

Topics: Introduction

Music, Brain, Emotions

- Psychological models of emotions and musical emotions 🎵
- Cognitive accounts: Musical structures 🤔
- Affective accounts: Musical expressions 😊
- Needs for “naturalistic” experiments 💁

Topics: Introduction

Music, Brain, Emotions

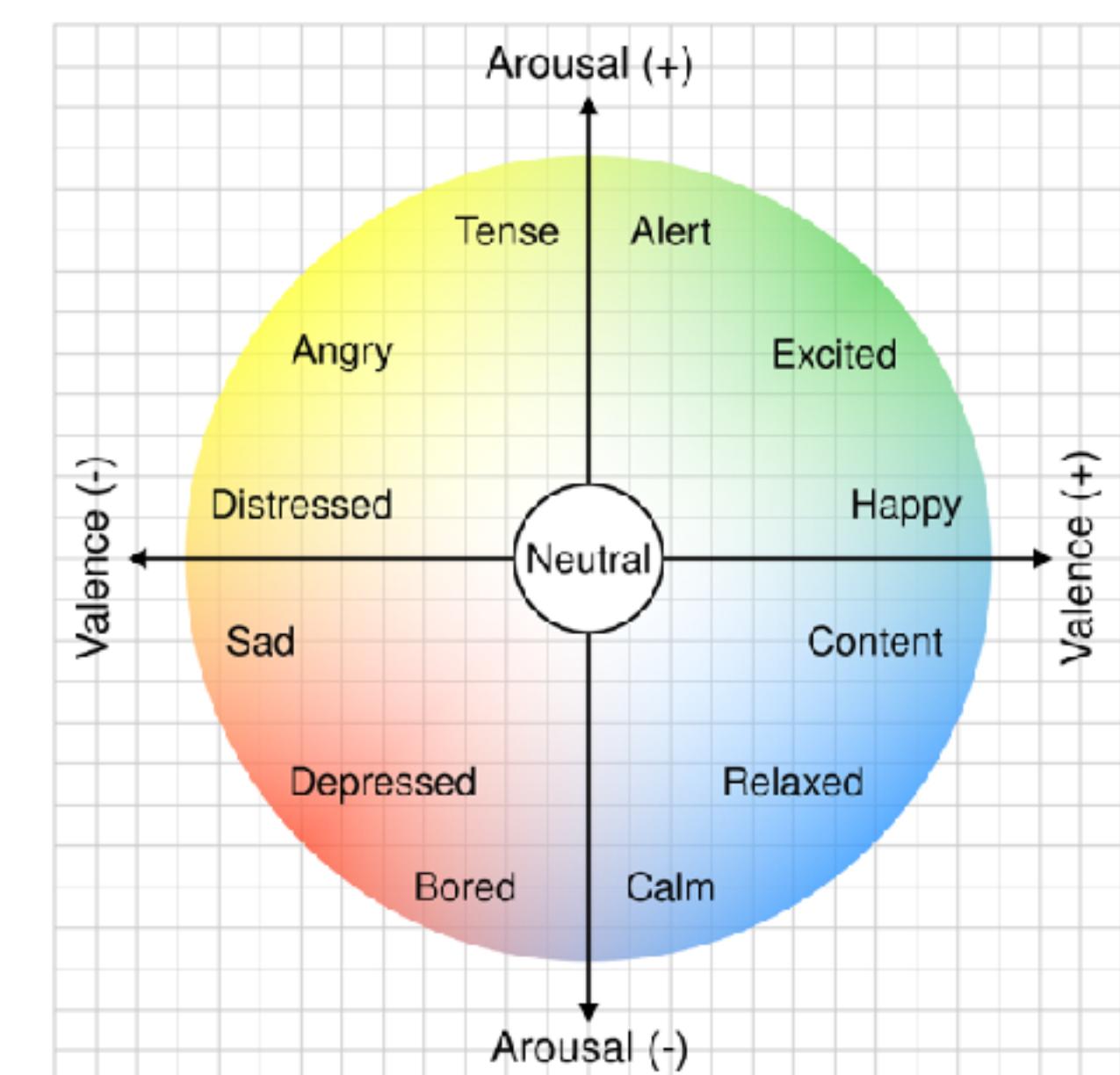
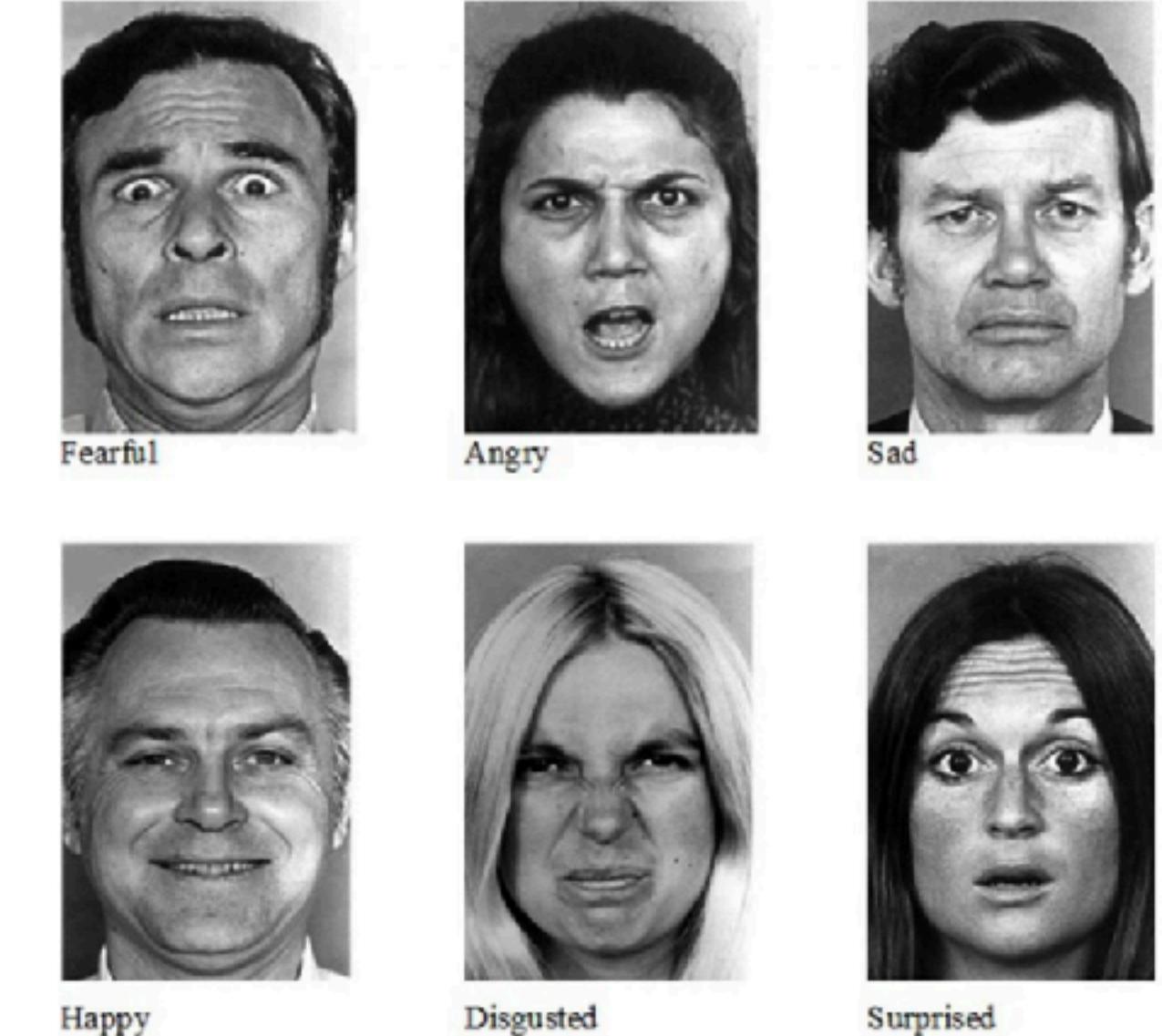
- **Psychological models of emotions and musical emotions** 🎵
- **Cognitive accounts: Musical structures** 🤔
- **Affective accounts: Musical expressions** 😊
- **Needs for “naturalistic” experiments** 🦋



Fundamental emotion theories

A brief overview

- **Physiological theories:**
 - James-Lange (→ 😰), Cannon-Bard (& 😰 = 😰)
 - Basic emotions: Ekman, K=6 (state = [😊, 😠, 😢, 😢, 😢, 😢]),, currently Cowen & Keltner, K=27
- **Cognitive (constructivist) theories:**
 - Schachter-Singer (→ 🧐 → 😰)
 - Cognitive appraisal: e.g., Lazarus, Scherer: core affect (e.g., [Valence, Arousal]) → 🧐 [semantics, goals, relevance, ...] → 😰



James Russel's circumplex model

Objective measures of subjective emotions

Why is it different from the qualia of loudness?

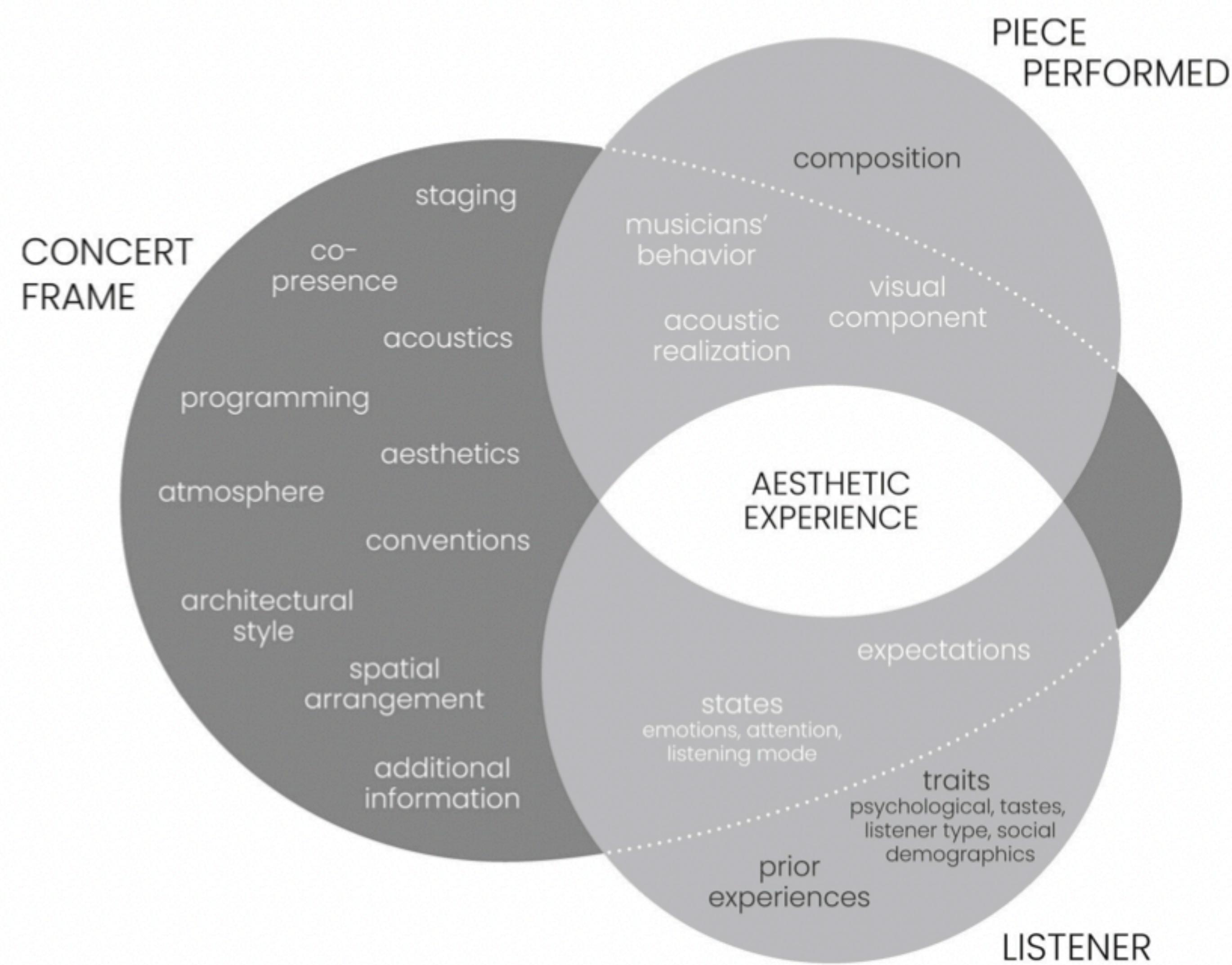
- **Why can't we just do psychophysics of emotions?** (e.g., 2AFC, odd-one-out, ...)
 - Because perceptual qualia is more consistent than emotions?
 - Because emotion is not a passive process? (constructivism)
 - (Bayesians: perception is not a passive process either!)
 - Or will the behavioral economics tell anything better about pleasure and values?
 - **What are your thoughts?** 

Do "aesthetic emotions" exist?

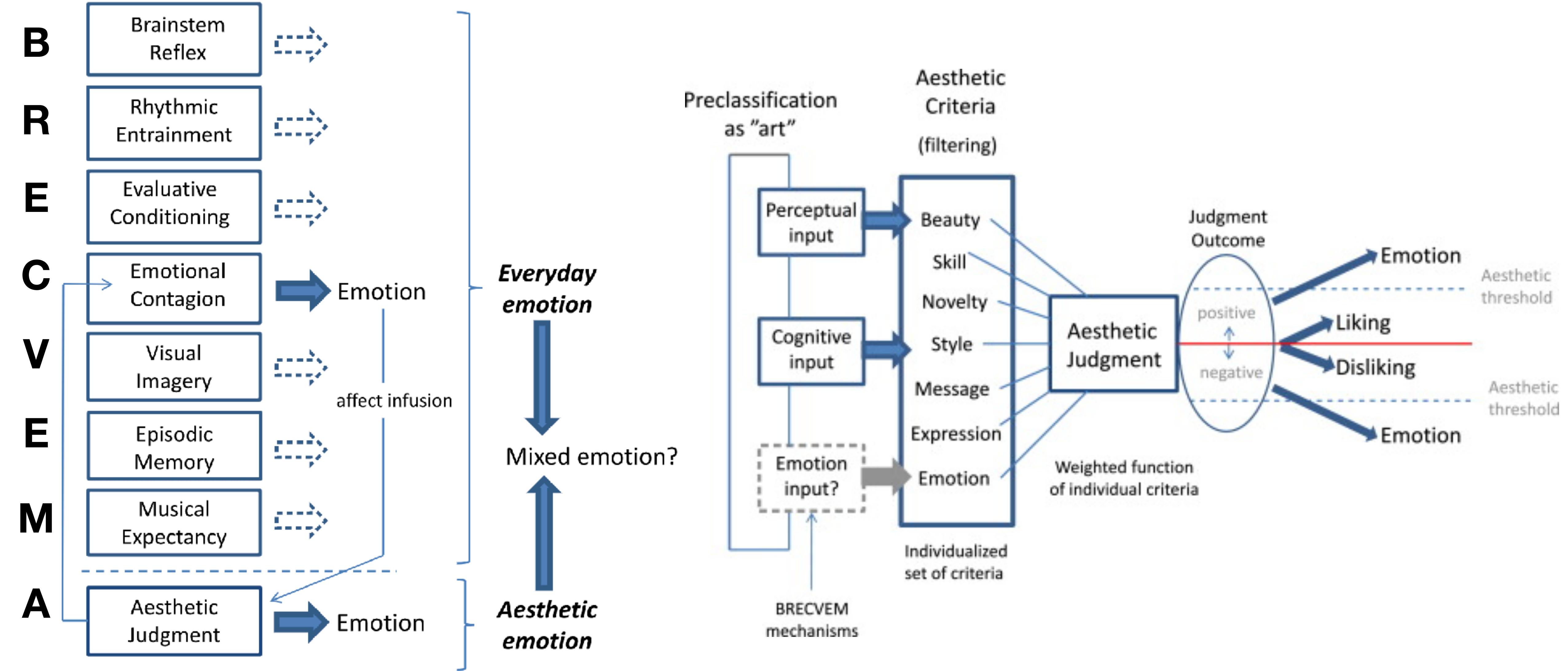
Menninghaus et al., 2019, 2020; Skov & Nadal, 2020, Psychological Rev

- **Menninghaus:** Yes, because Kant said so ("disinterested pleasure"). Ambivalent emotions (e.g., bittersweet, pleasurable melancholy) are more common than everyday's emotions.
- **Skov:** No, because the affective states during aesthetic appreciation are not distinctive from other sensory valuations.
- **Menninghaus:** Still yes, because aesthetic emotions is essential for aesthetic evaluation. (hmm...🤔)

Theories on musical emotions



Theories on musical emotions



Topics: Introduction

Music, Brain, Emotions

- Psychological models of emotions and musical emotions 🎵
- Cognitive accounts: Musical structures 🤔
- Affective accounts: Musical expressions 😊
- Needs for “naturalistic” experiments 💁

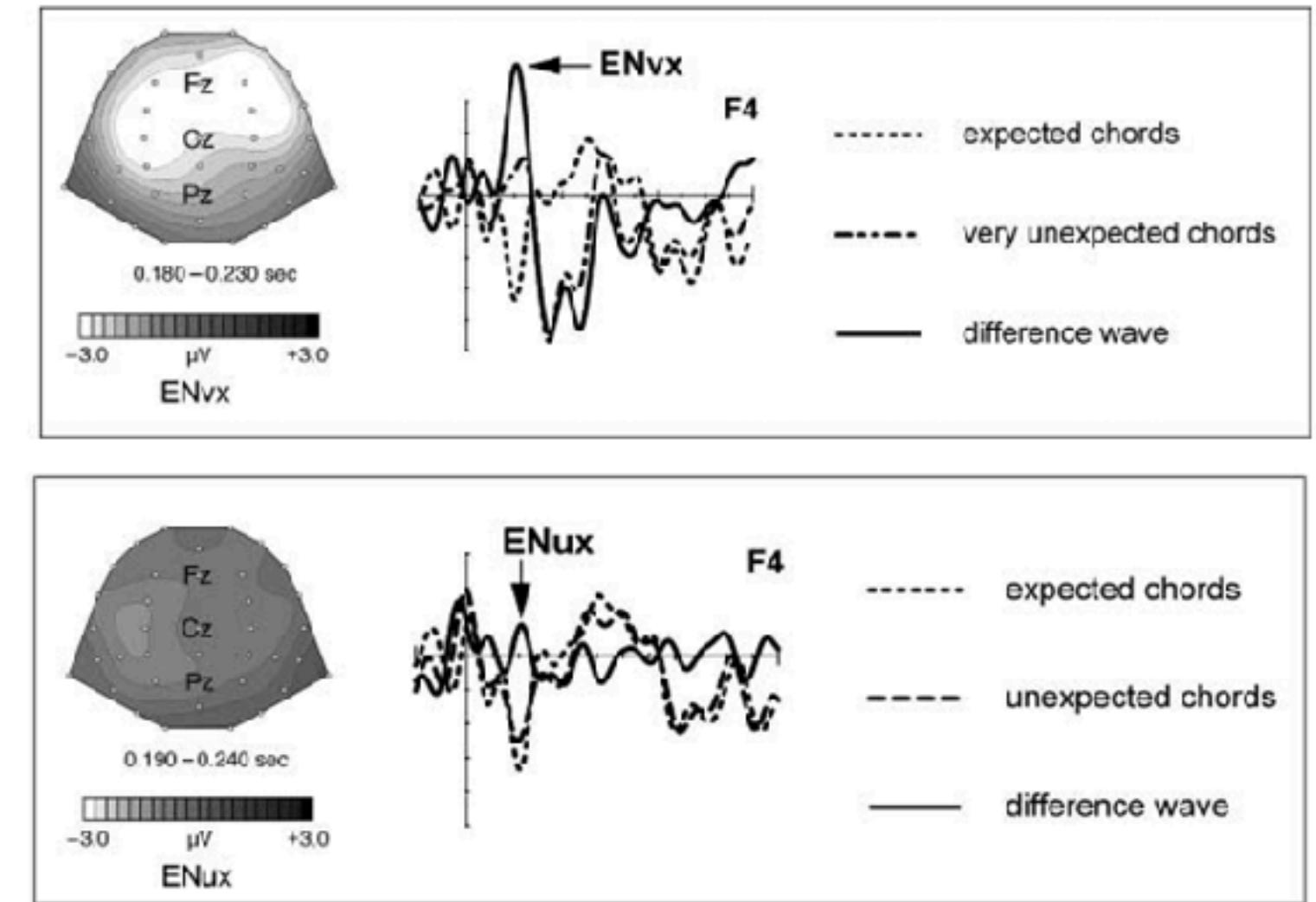
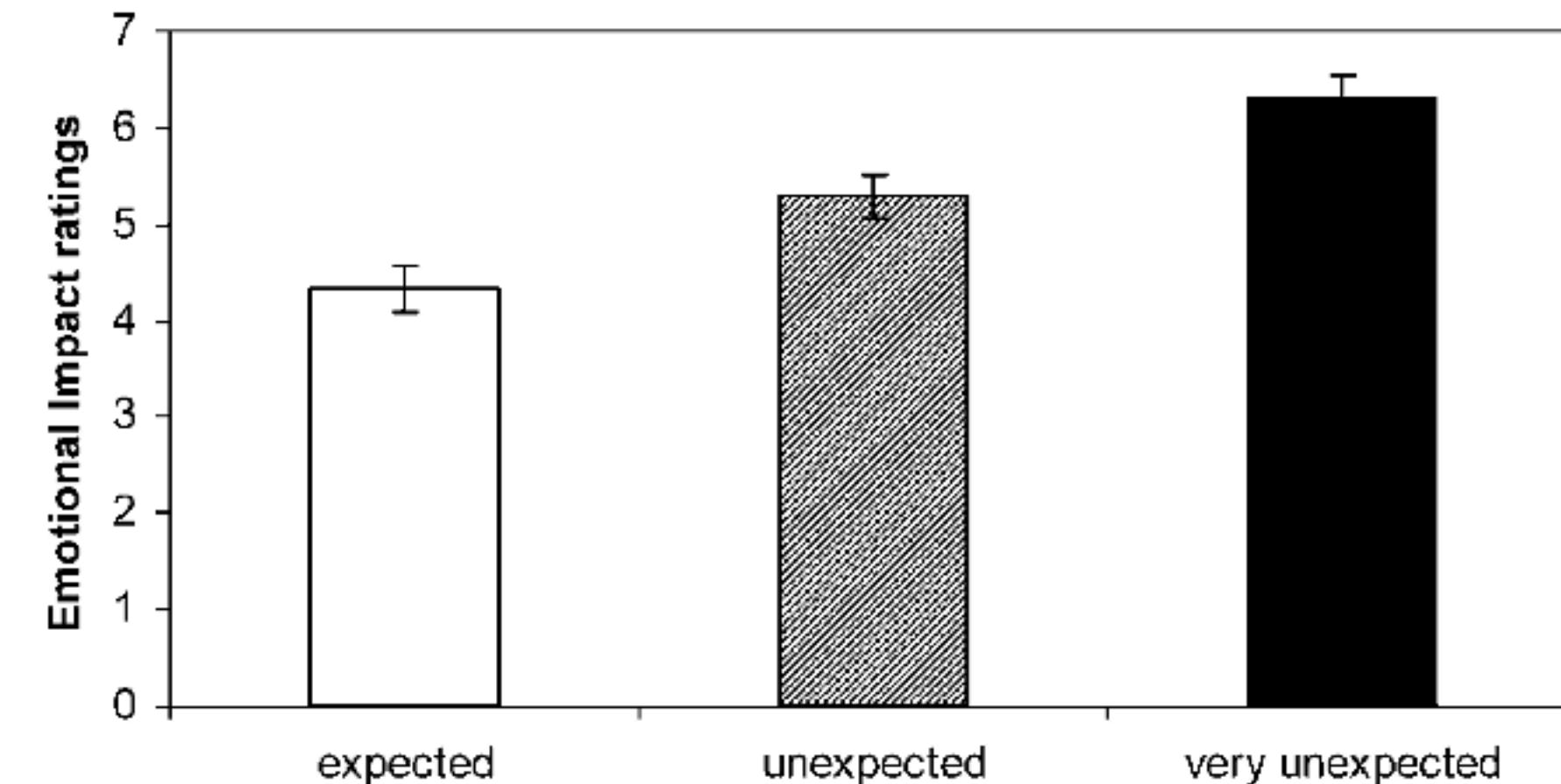
Cognitive accounts

Listeners (even implicitly) perceives musical elements.

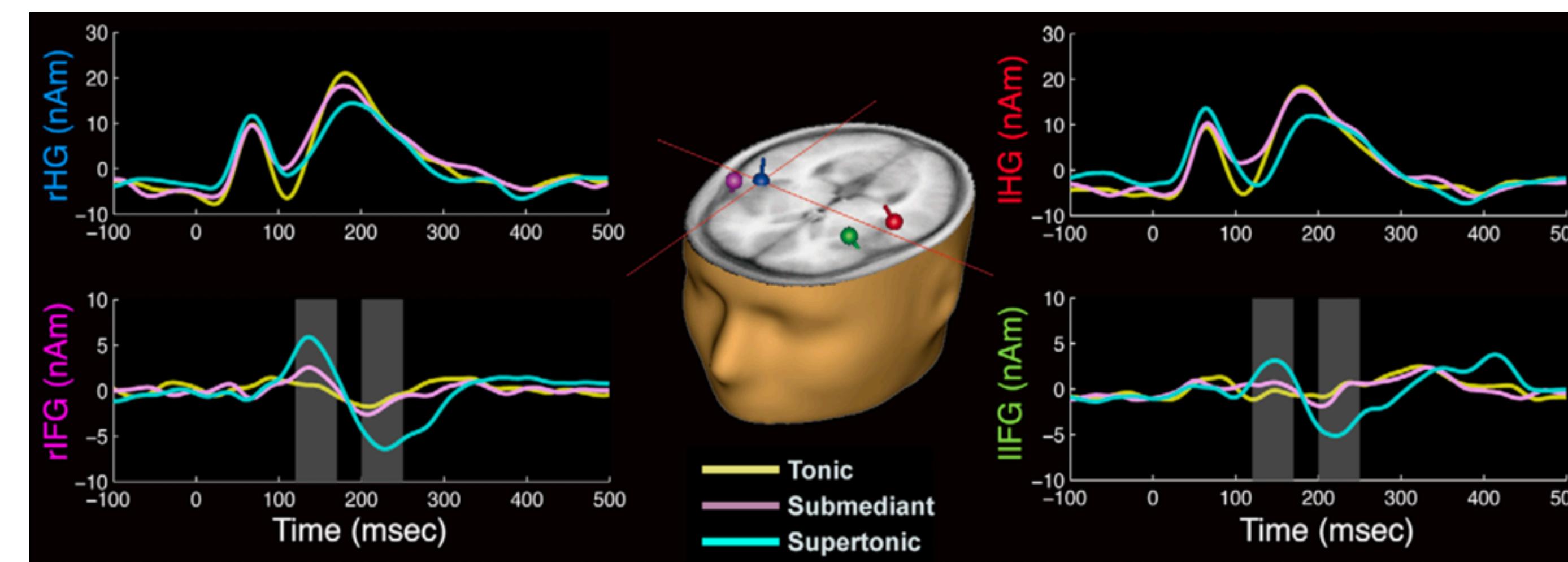
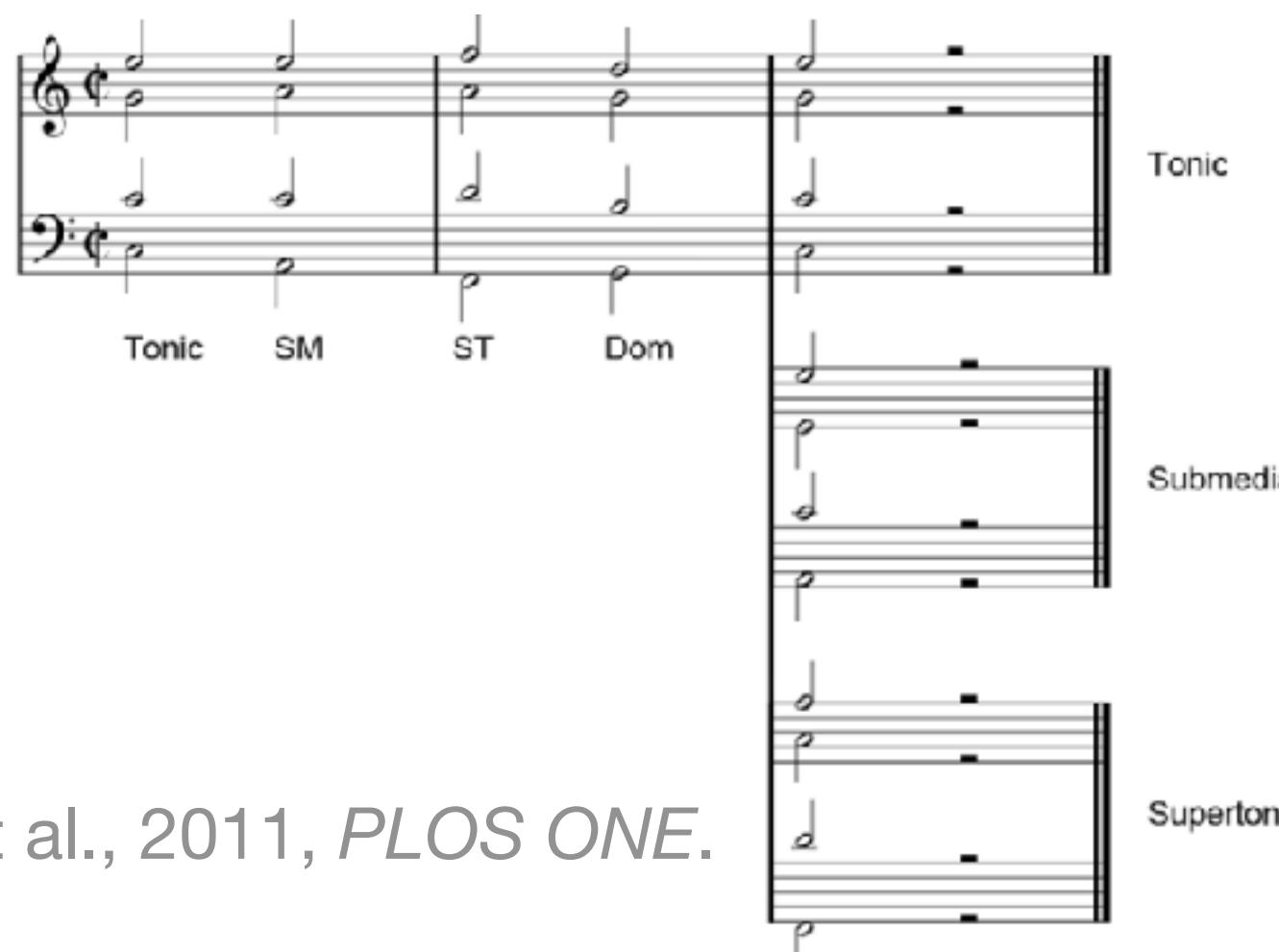
- **Spectral:** tonality (keys, scales, modes), pitch, intervals, chords, ...
- **Temporal:** meters, beats, rhythms, ...
- **Interplays of complexity** of patterns, forming and betraying **expectations**, leading to "Tension and Resolution" (Meyer, 1956)

Cognitive processing of music

E.g., Tonal perception and expectations

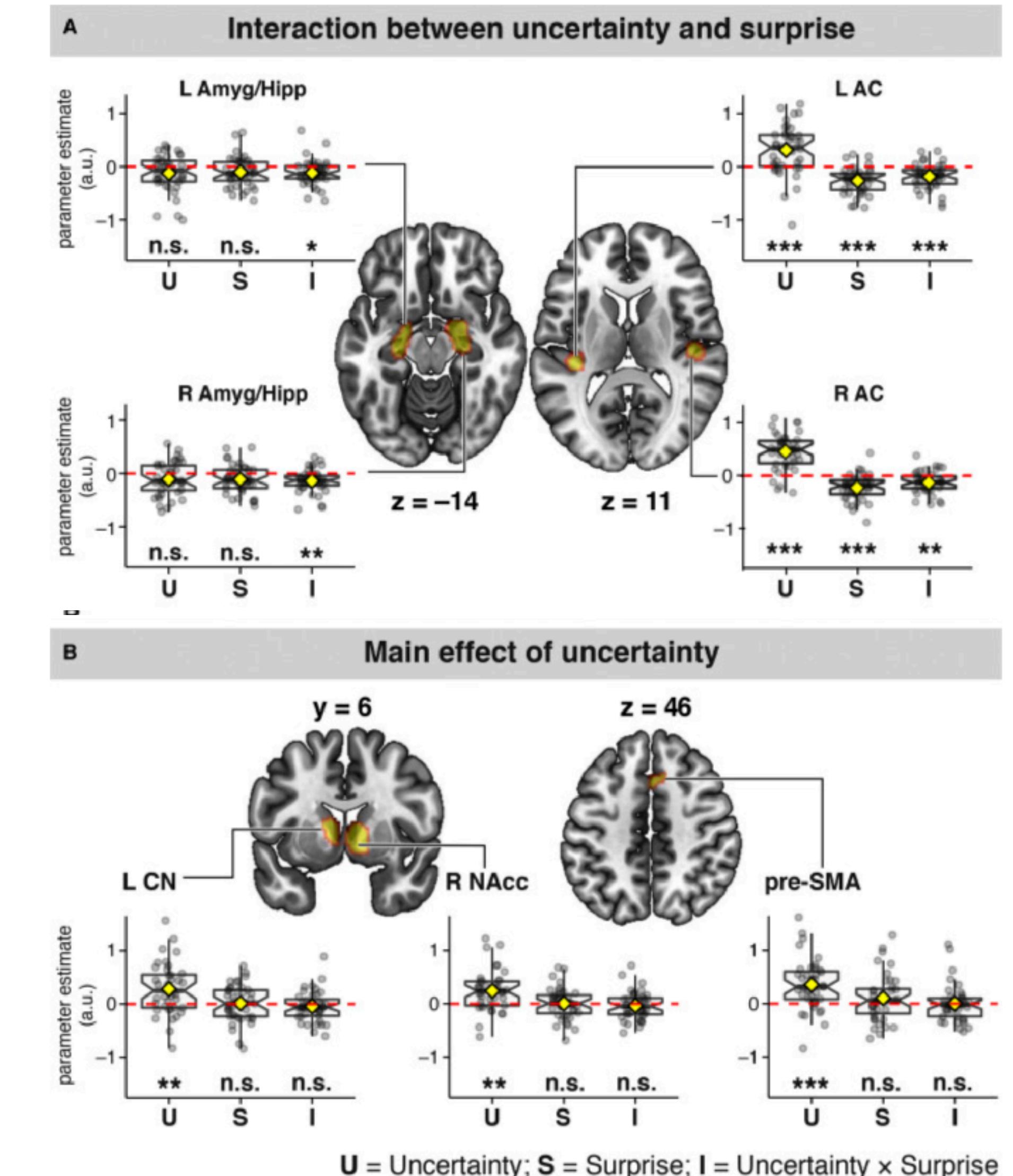
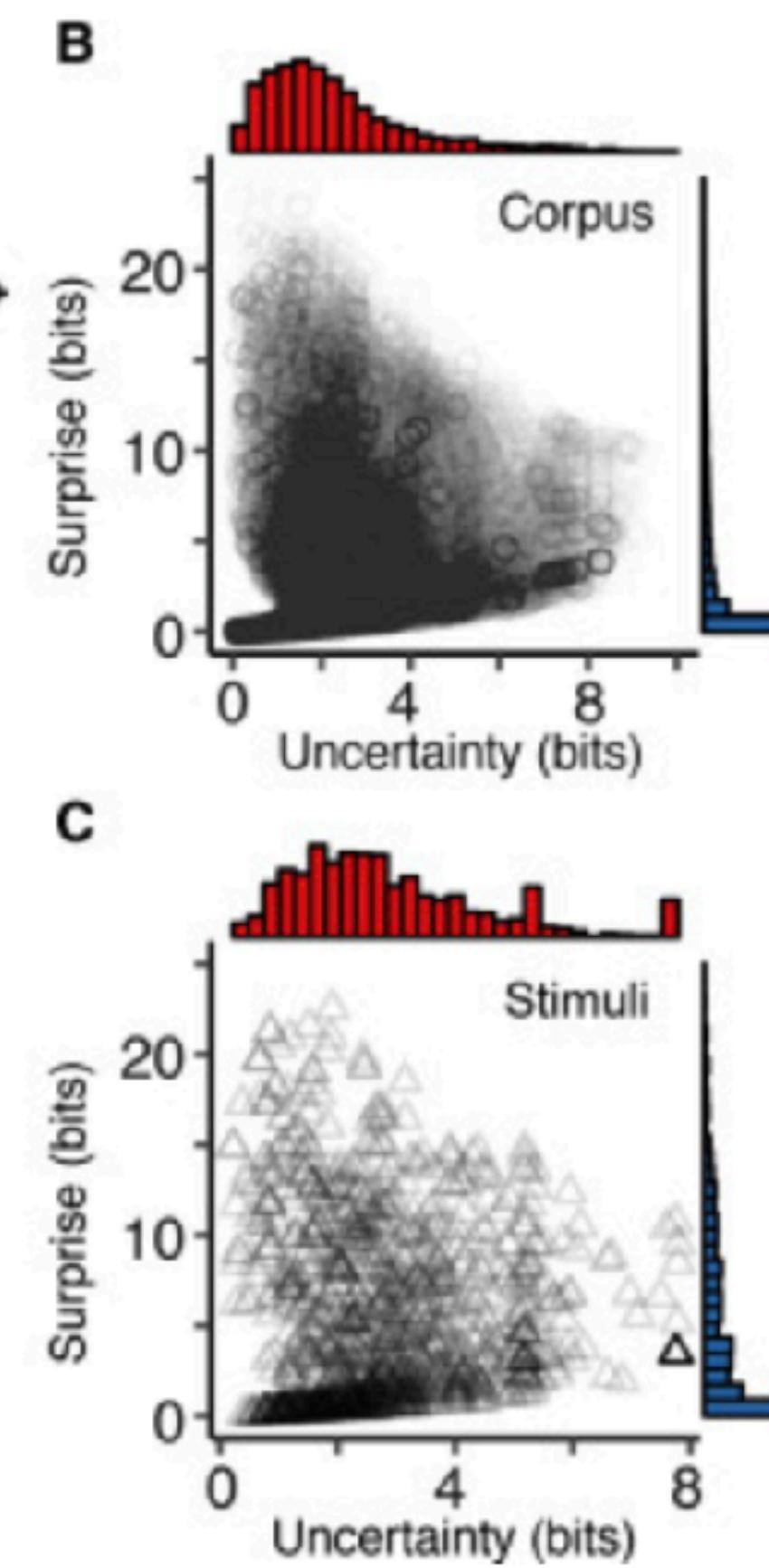
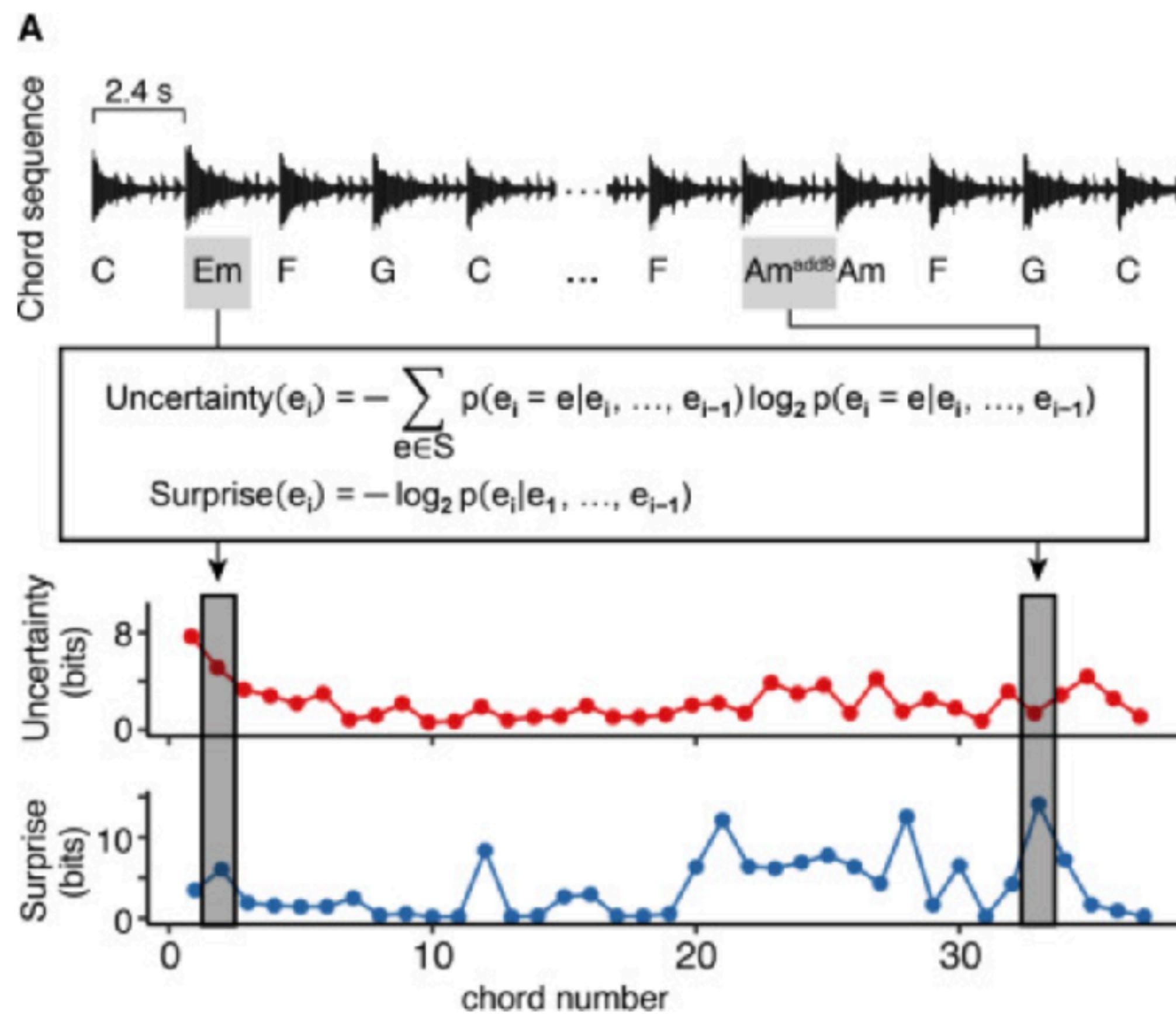


Steinbeis, Sloboda, Koelsch, 2006, *J Cognitive Neuroscience*.



Cognitive processing of music

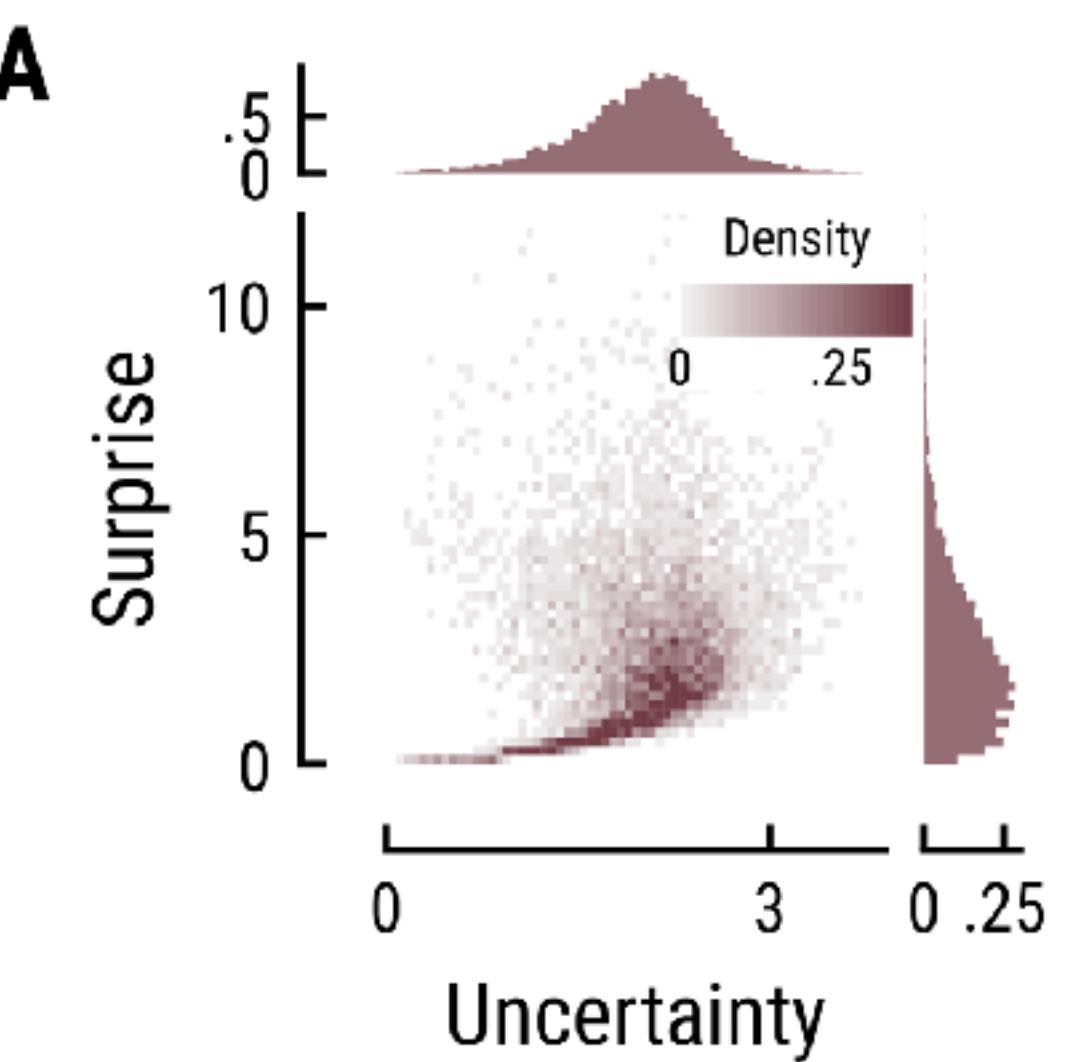
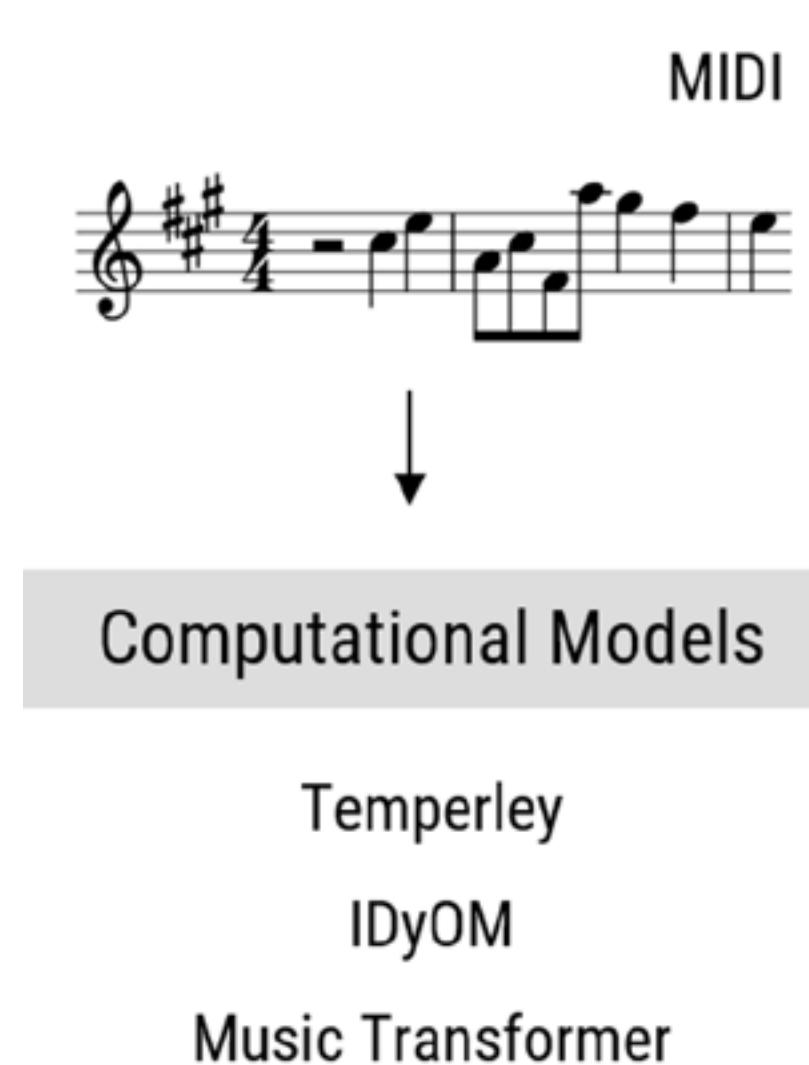
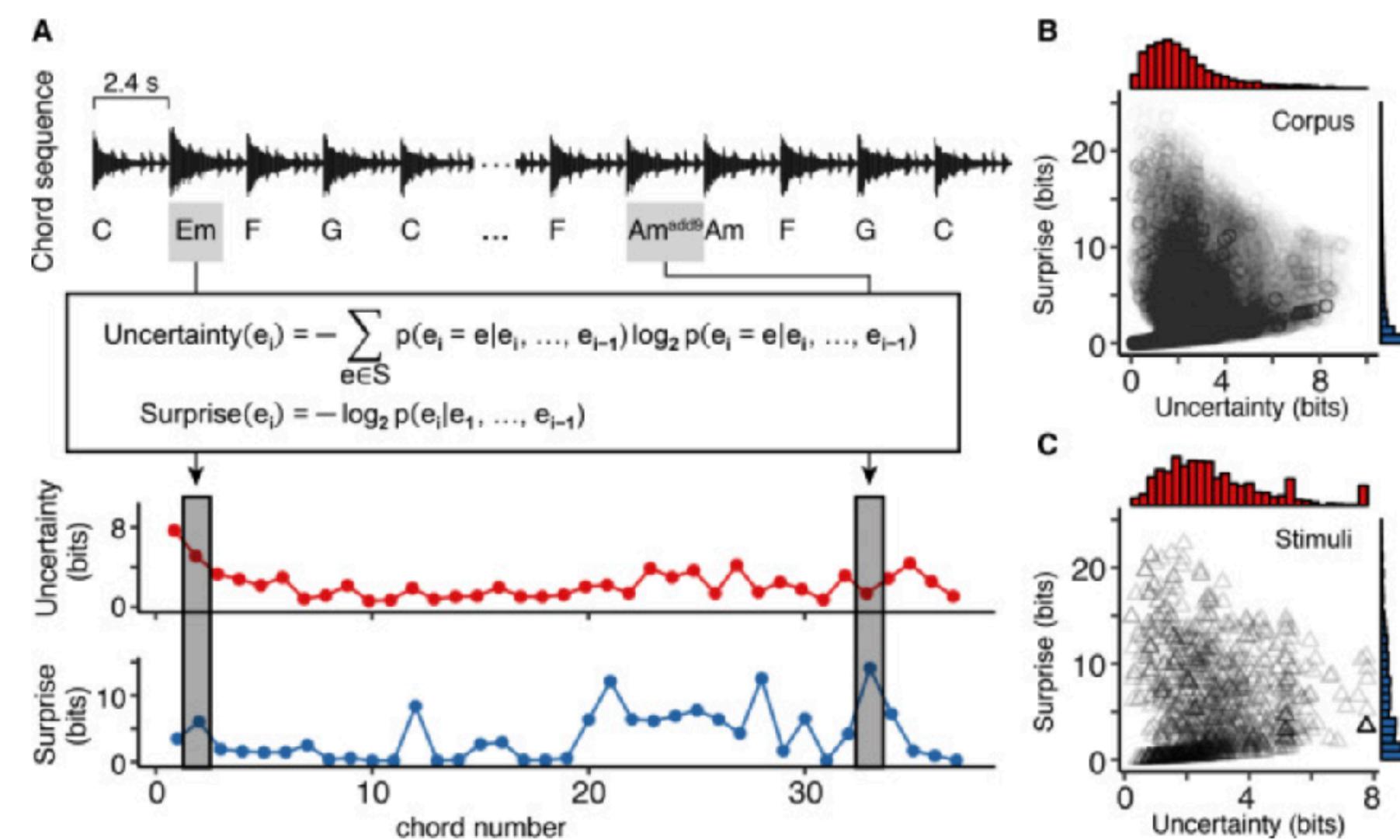
E.g., Tonal perception and expectations



Symbol-domain models

Input: notes on the score

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 symbolic models often disregards expressions.

SOUND DEMO: "Unkempt Hair (쑥대머리)", Chun-hyang-ga (춘향가), Juri Kim (김주리), 2023

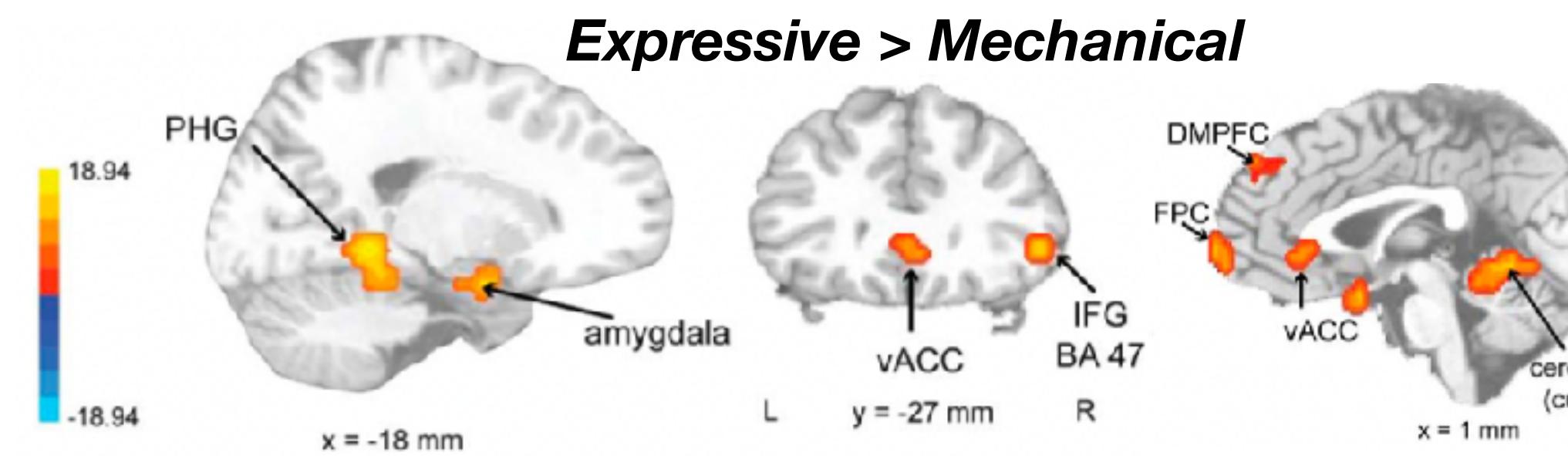
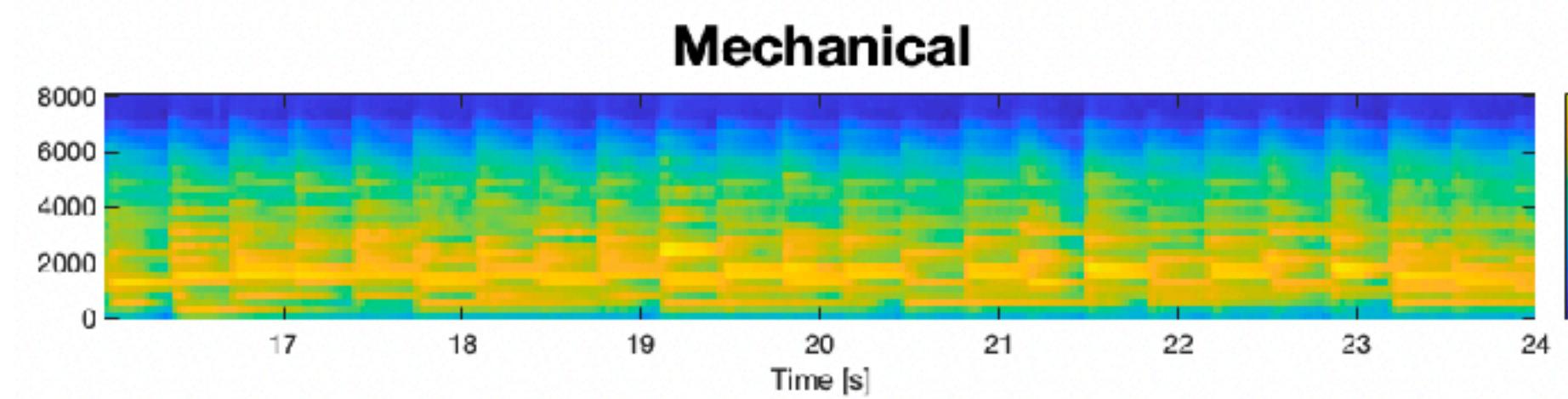
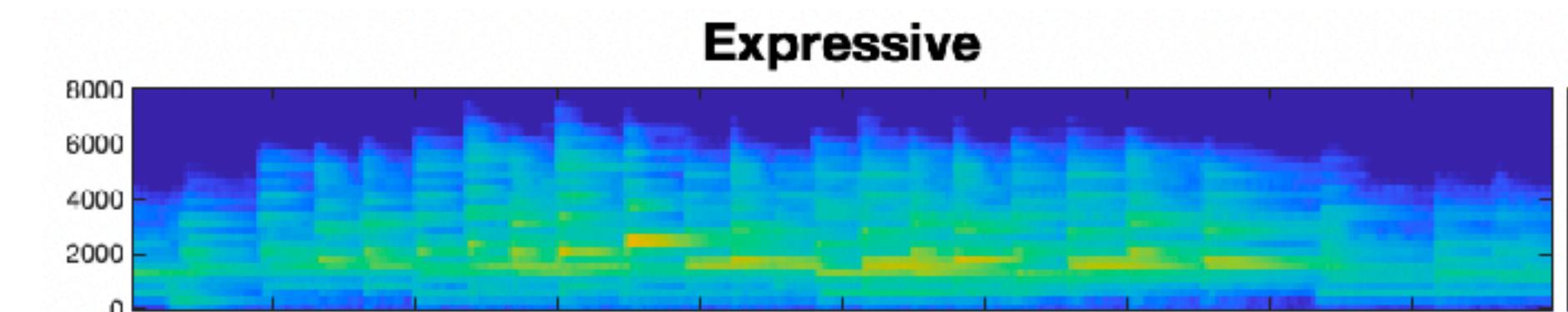
Topics: Introduction

Music, Brain, Emotions

- Psychological models of emotions and musical emotions 🎵
- Cognitive accounts: Musical structures 🤔
- Affective accounts: Musical expressions 😊
- Needs for “naturalistic” experiments 💁

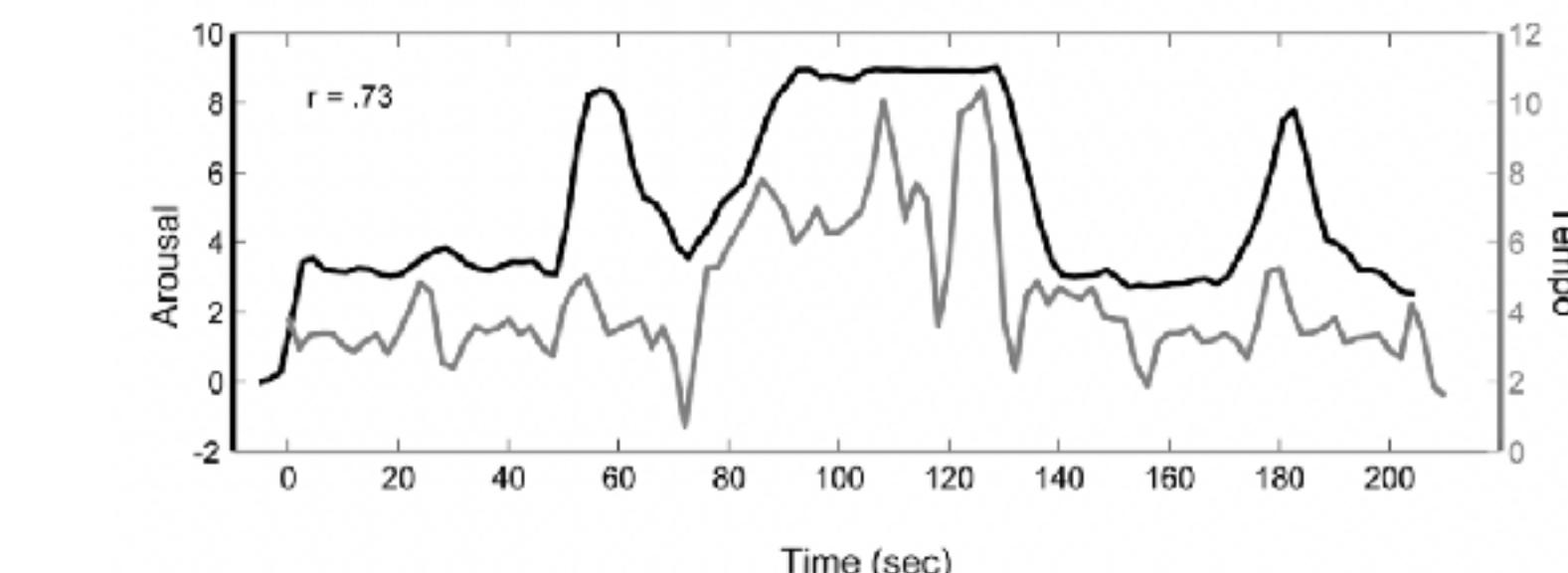
Cognitive processing of music

E.g., local dynamics and tempi perturbations

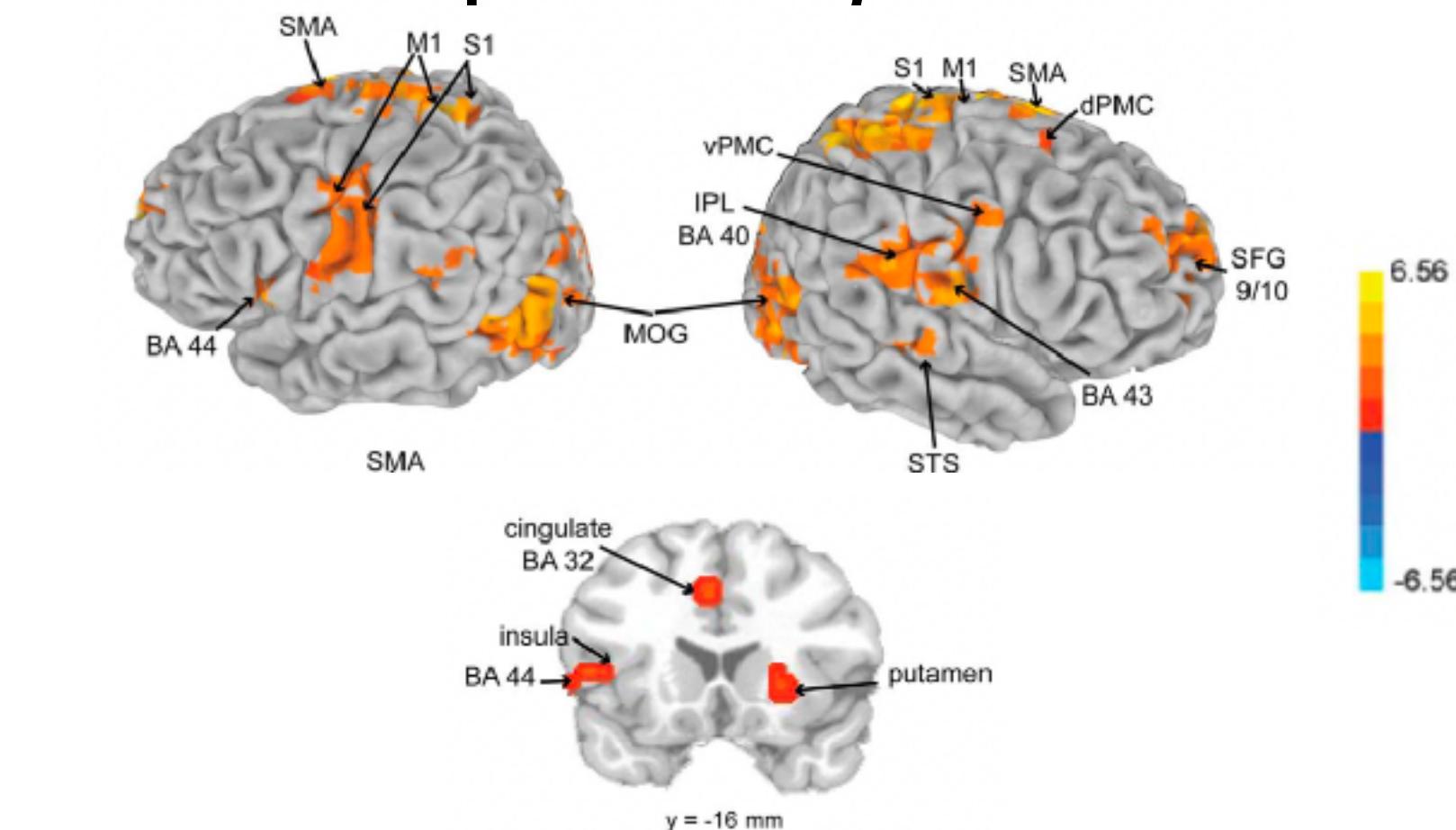


Chapin et al., 2010, PLOS One.

An exemplary participant (lag = 5.4 s)



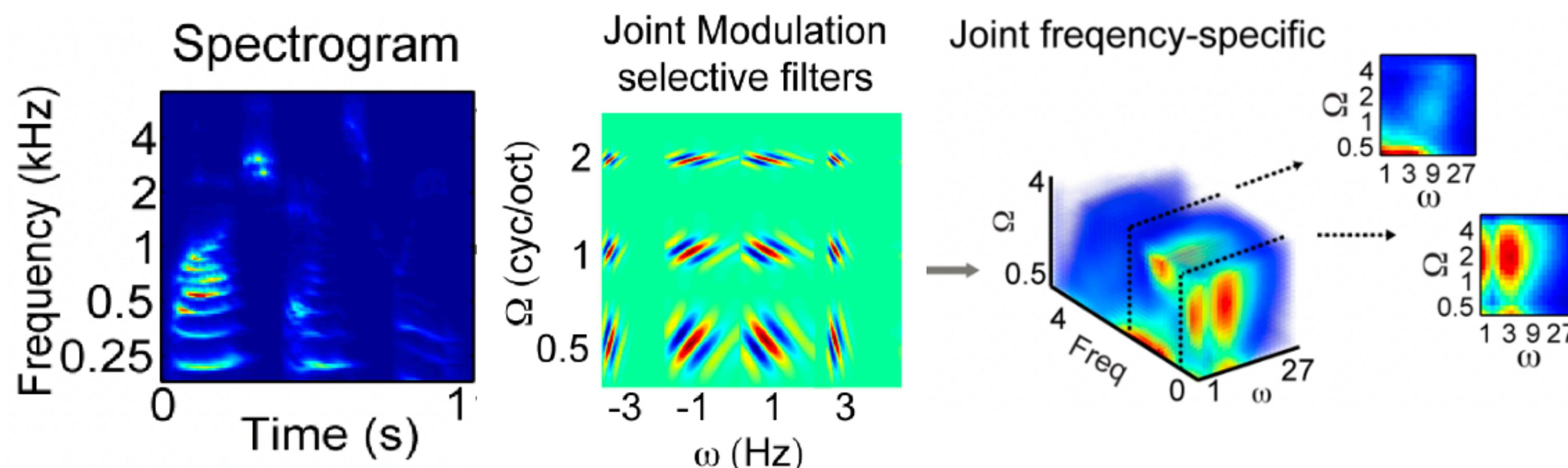
Tempo within Expressive



Audio-domain models

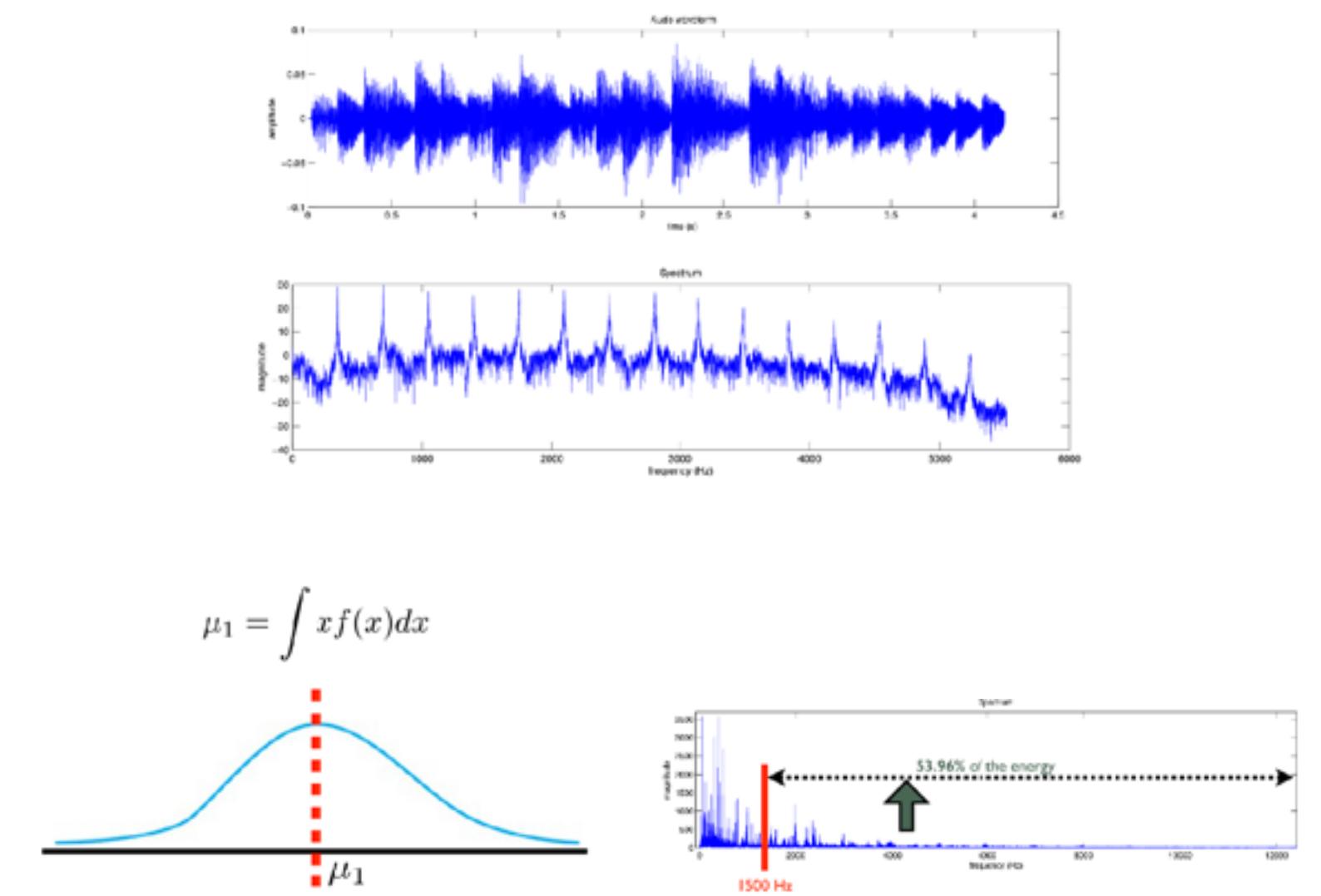
Input: audio samples

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

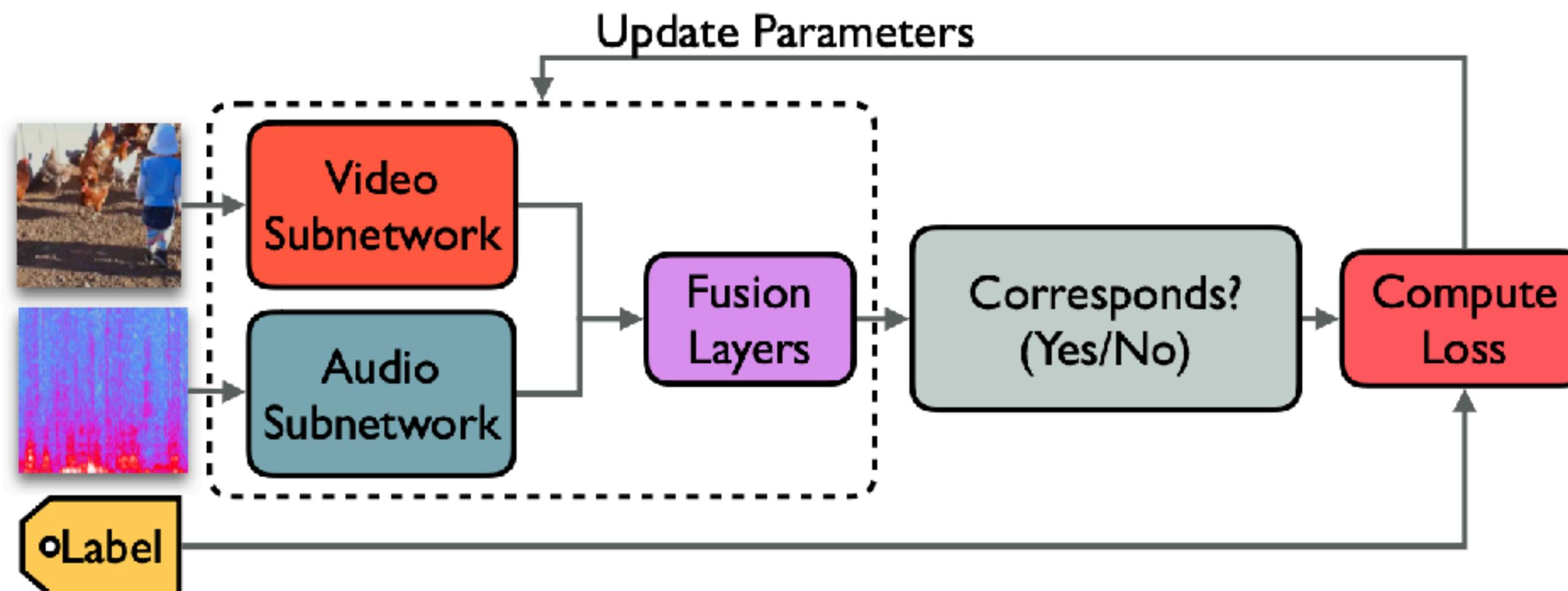


Lartillot, 2017, *MIRtoolbox User's Manual*.

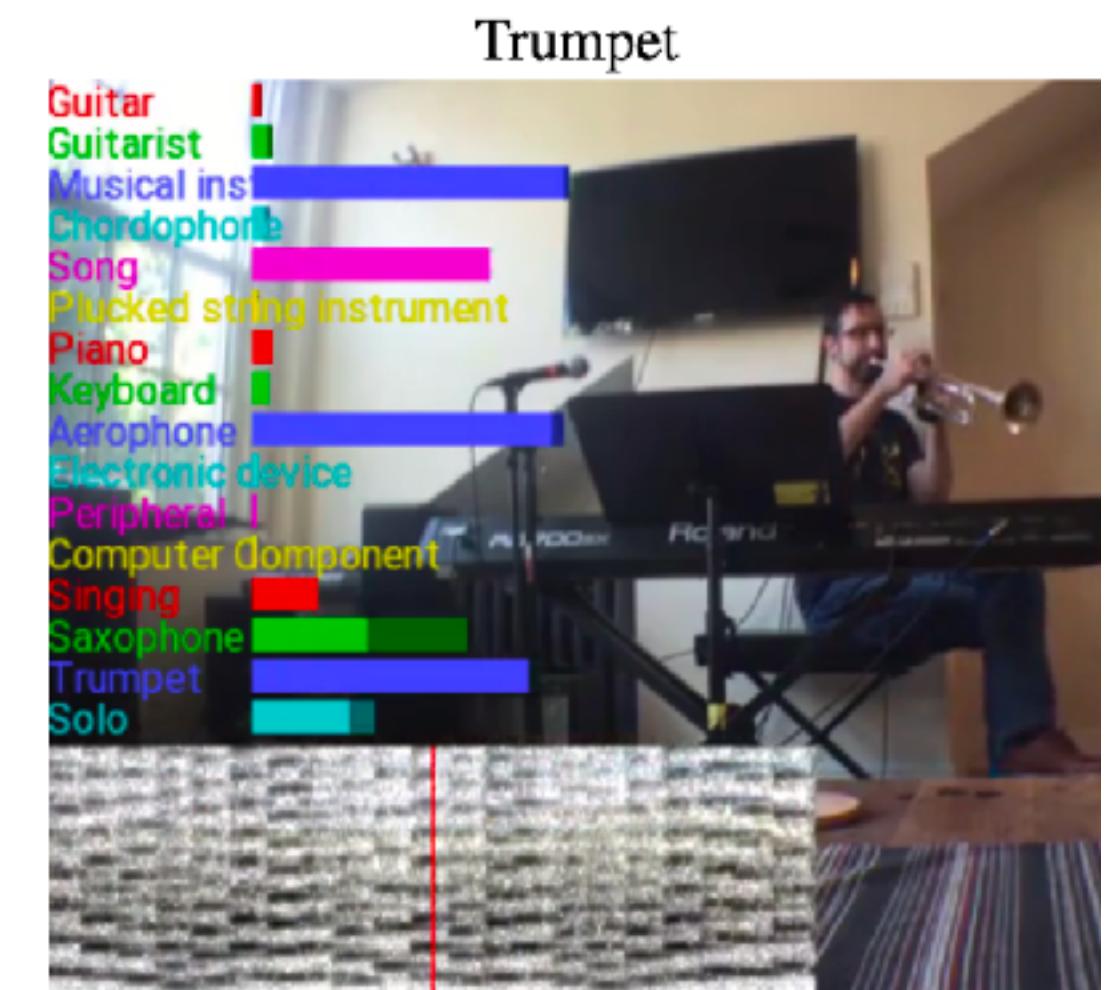
Traditional models mostly simulate cochlear and primary auditory cortex activity.

"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** (e.g., "this kind of spectrogram often goes with that video").



Open-L3: Cramer et al., 2019, ICASSP



VGGish: Hershey et al., 2017, ICASSP

A possible mid/high-level representation of sounds

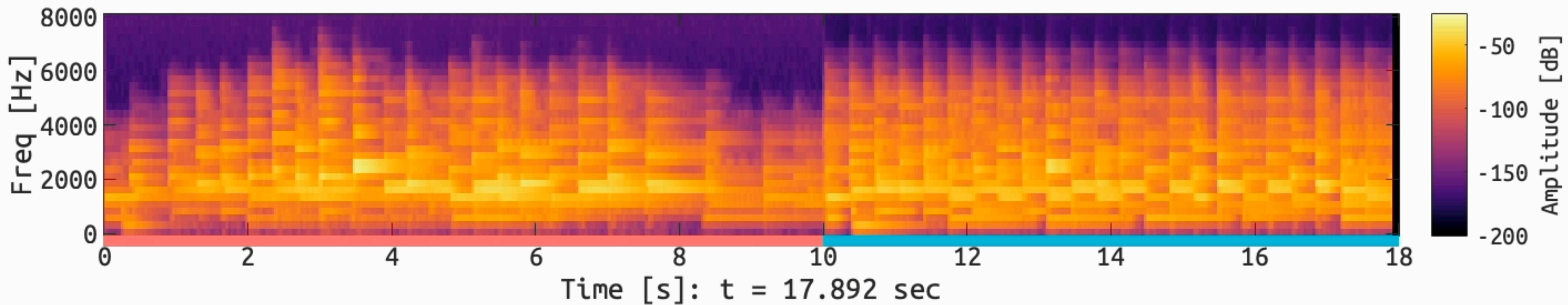
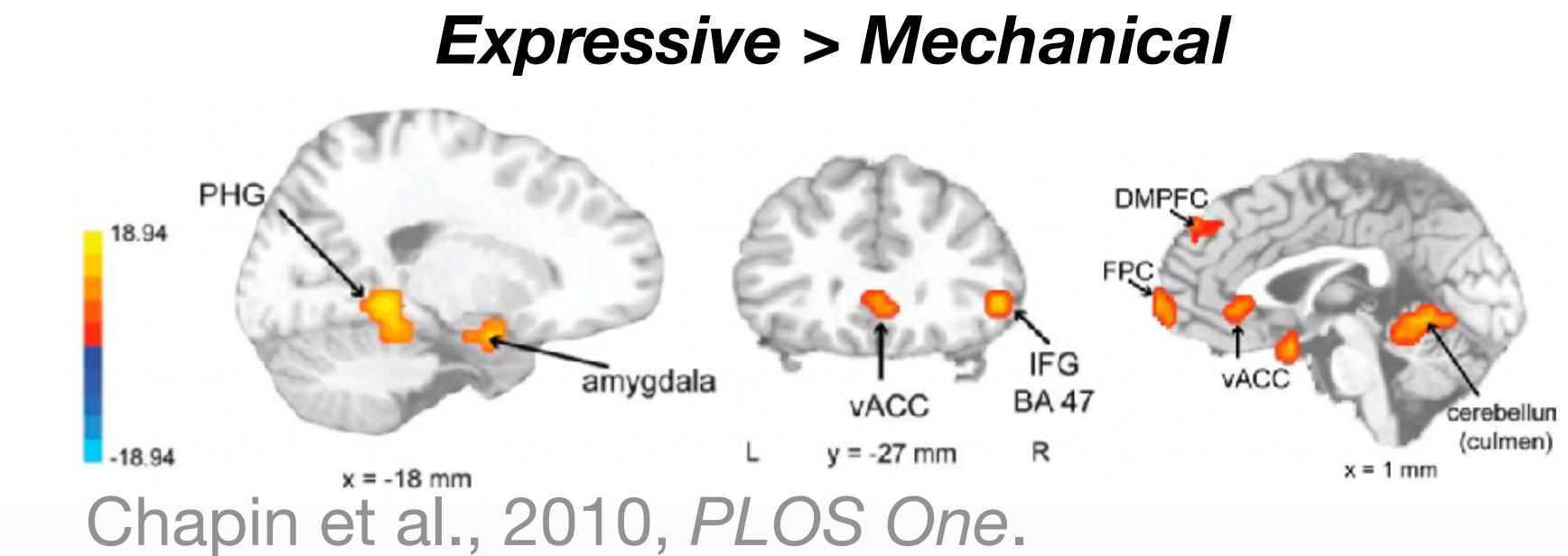


**But is it really expression?
Isn't it just timbre?**

Example: Expressive vs. Mechanical

Chapin et al., 2010, PLOS ONE.

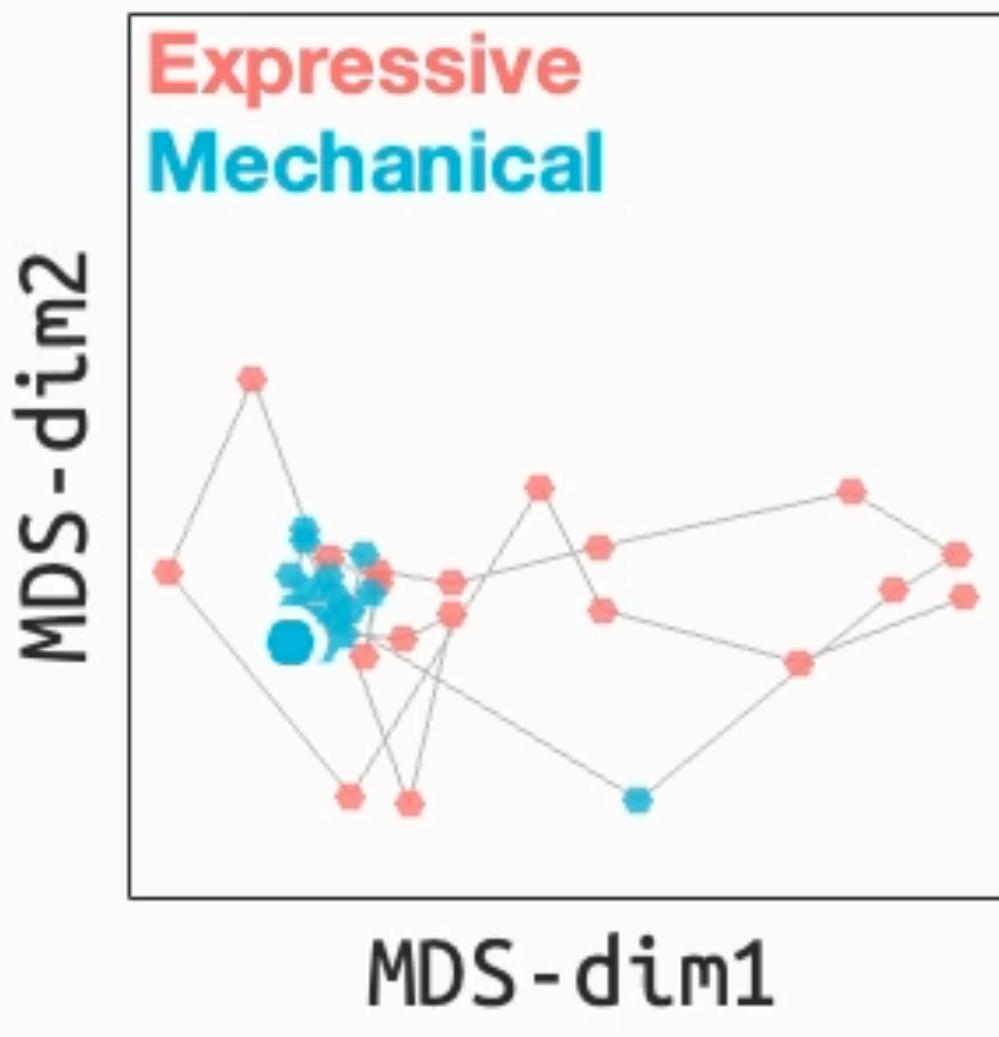
[EXPRESSIVE]: "Frédéric Chopin's Etude in E major, Op.10, No. 3 was performed by an undergraduate piano major (female, 22 years old) on a Kawai CA 950 digital piano"



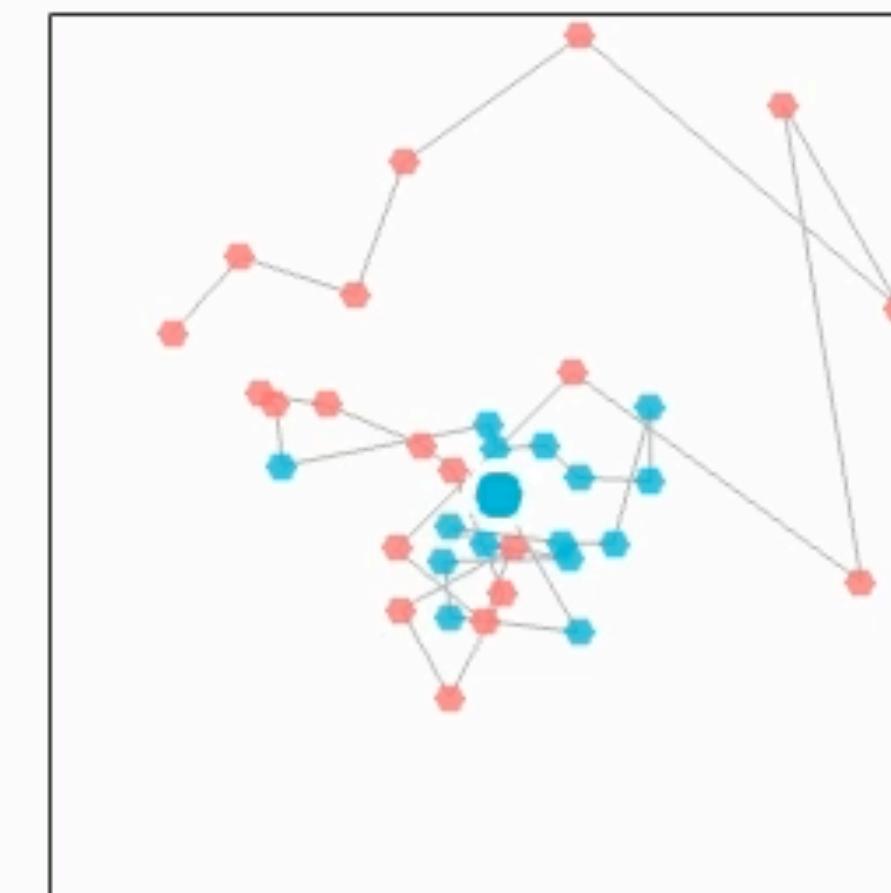
[MECHANICAL]: "The MIDI (Musical Instrument Digital Interface) onset velocity (key pressure) of each note (correlating with sound level) was set to 64 (range 0–128), and pedal information was eliminated."

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

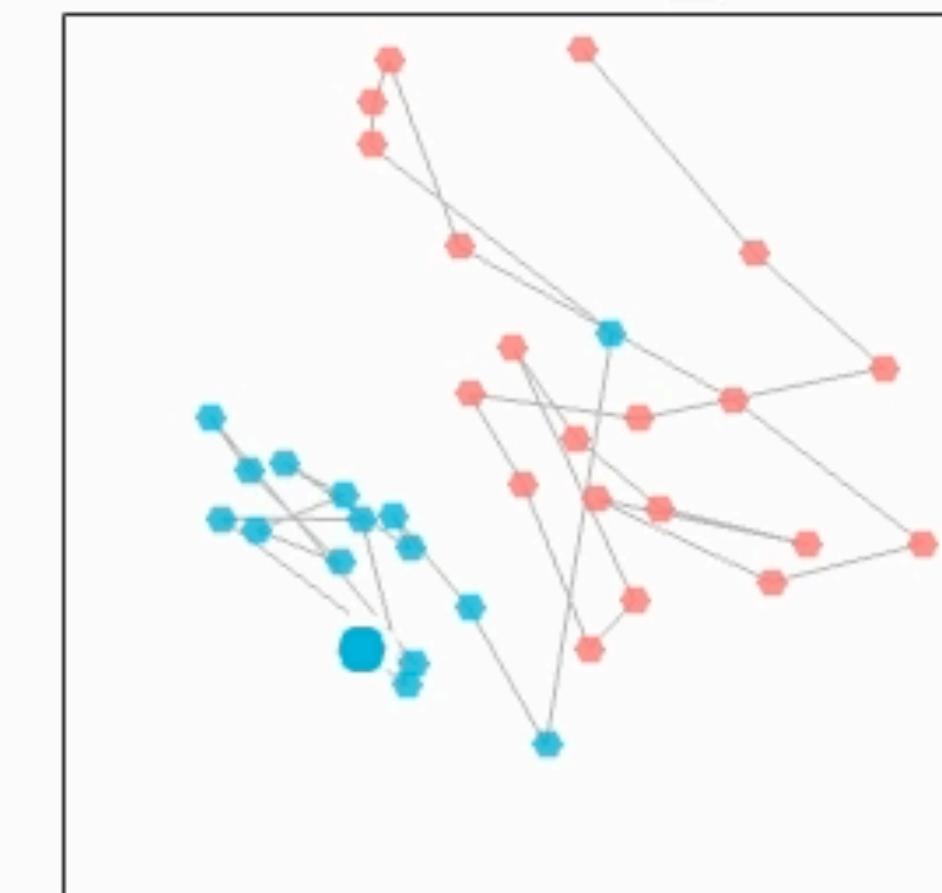
L01: InputBatch



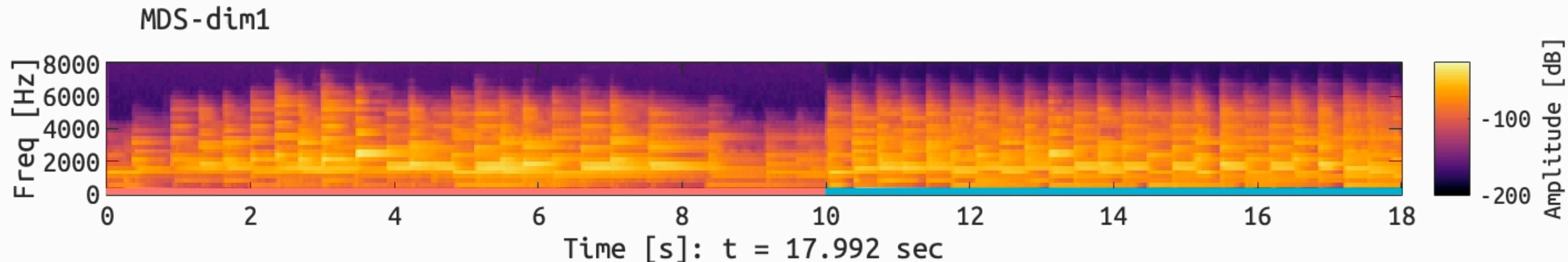
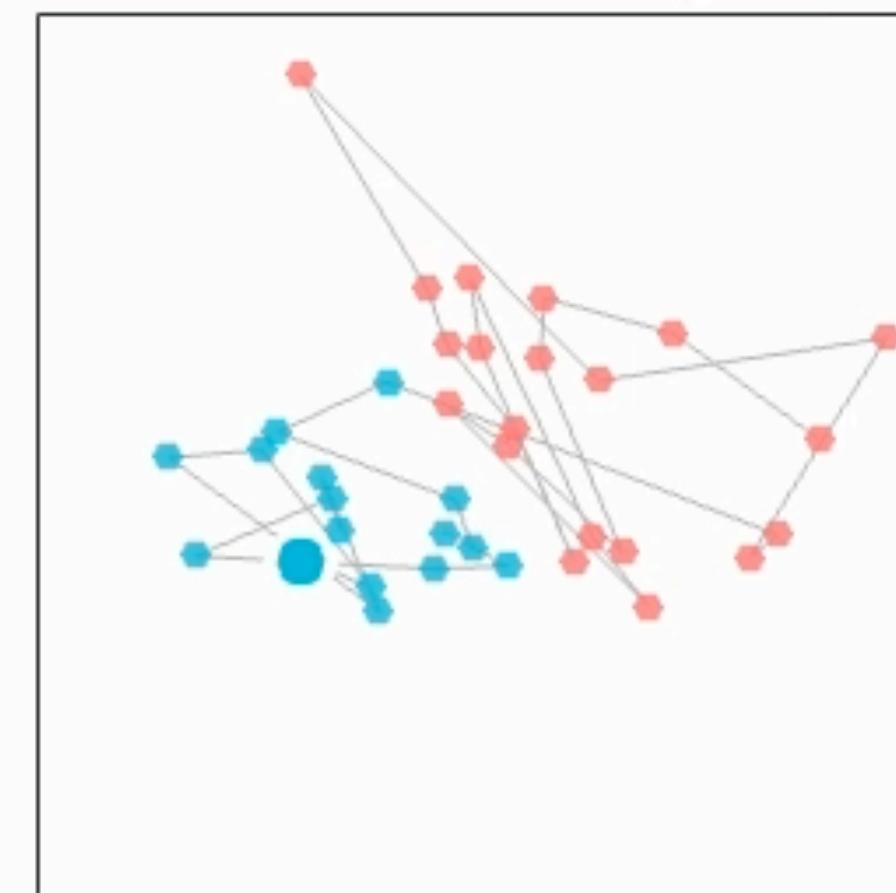
L02: conv1



L18: fc1_1



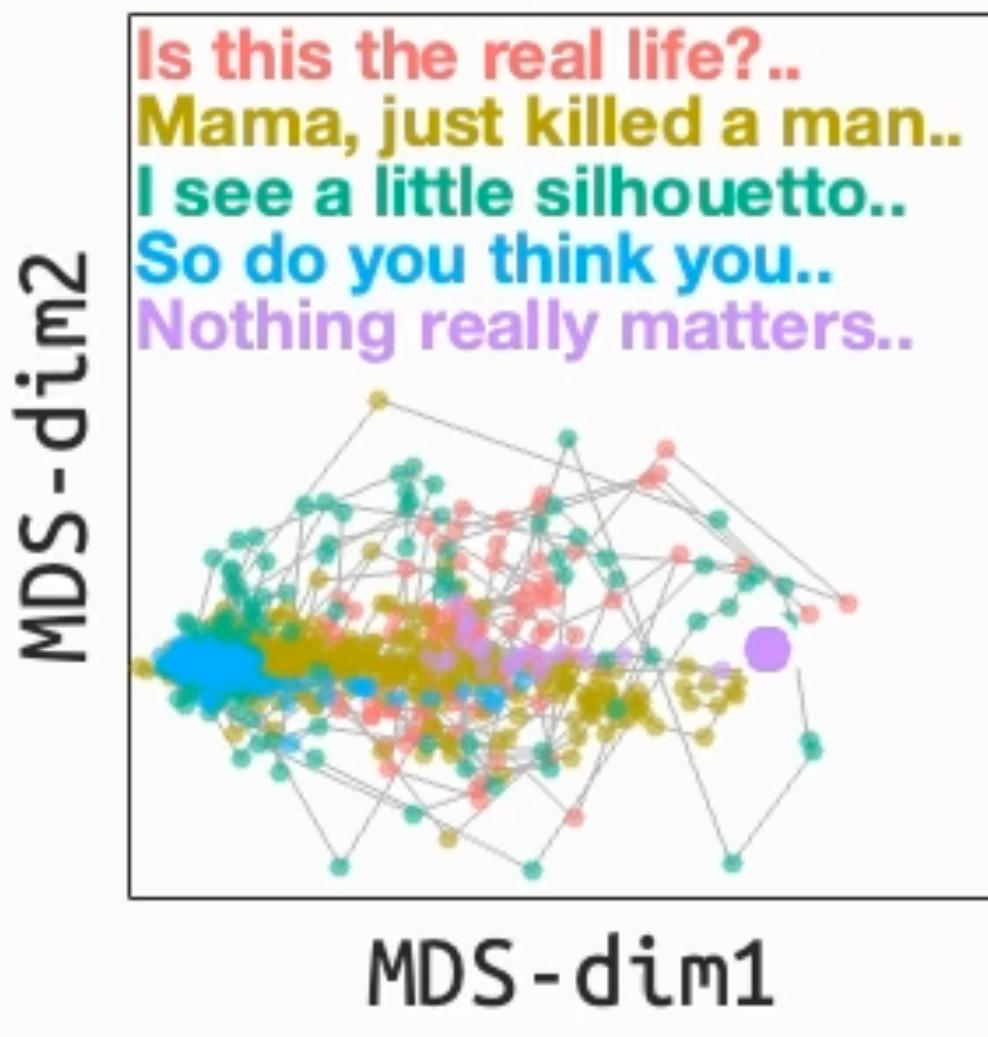
L23: EmbeddingBatch



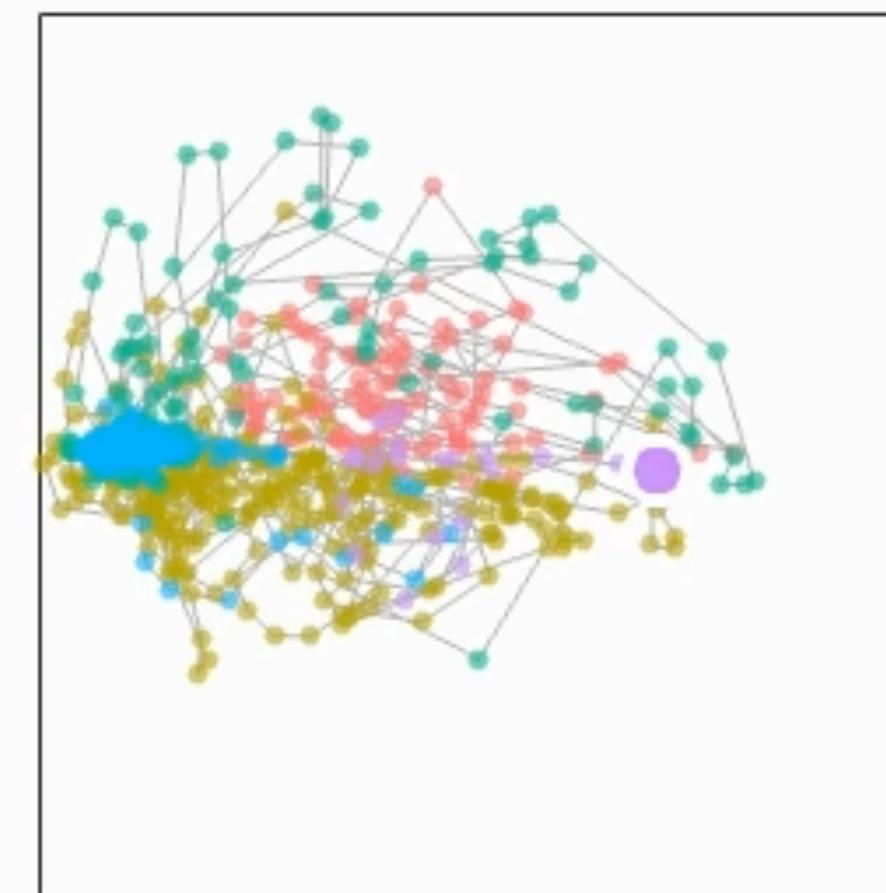
Example: Queen. (1975). Bohemian Rhapsody

VGGish representations at various layers (from 1 to 23)

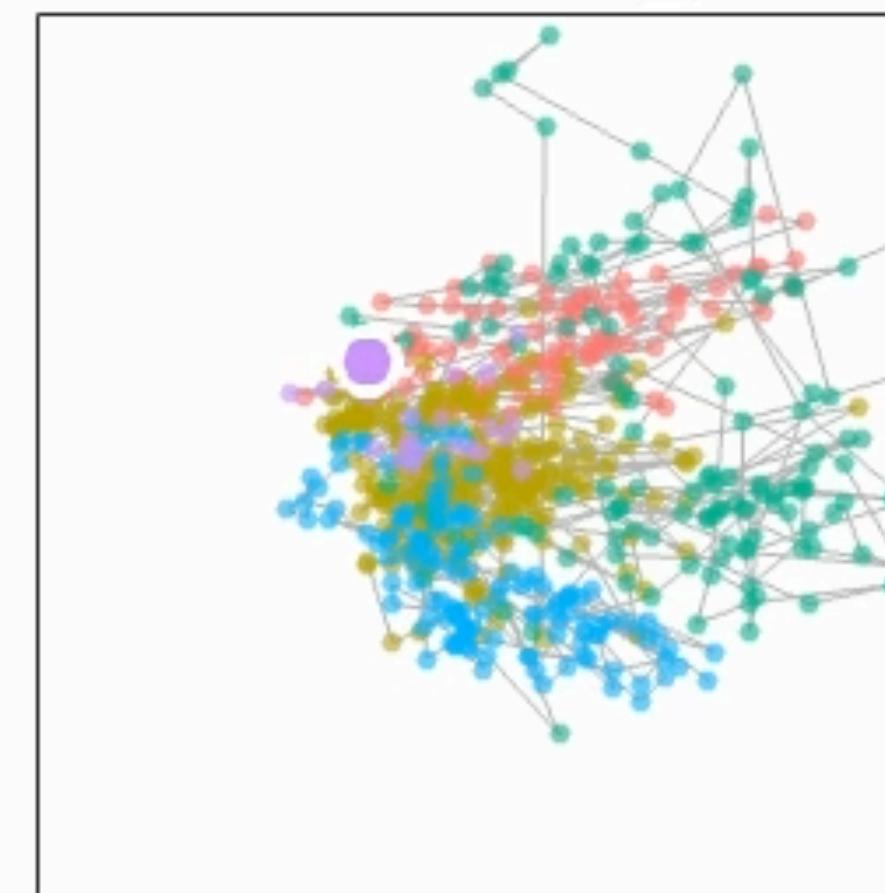
L01:InputBatch



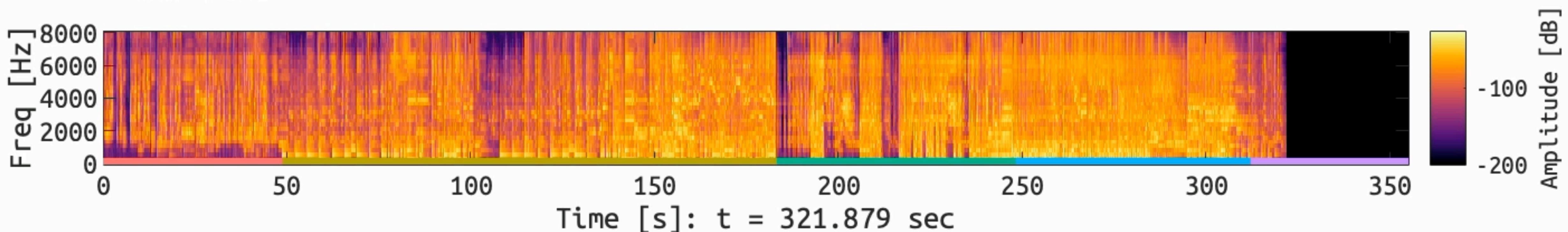
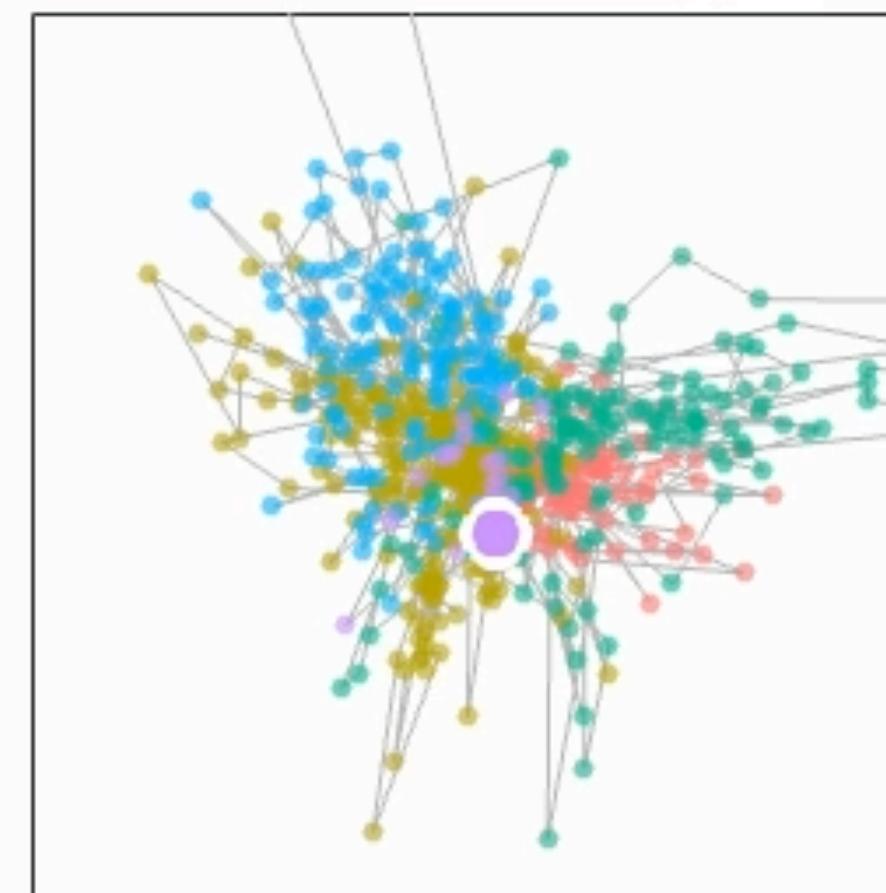
L02:conv1



L18:fc1_1



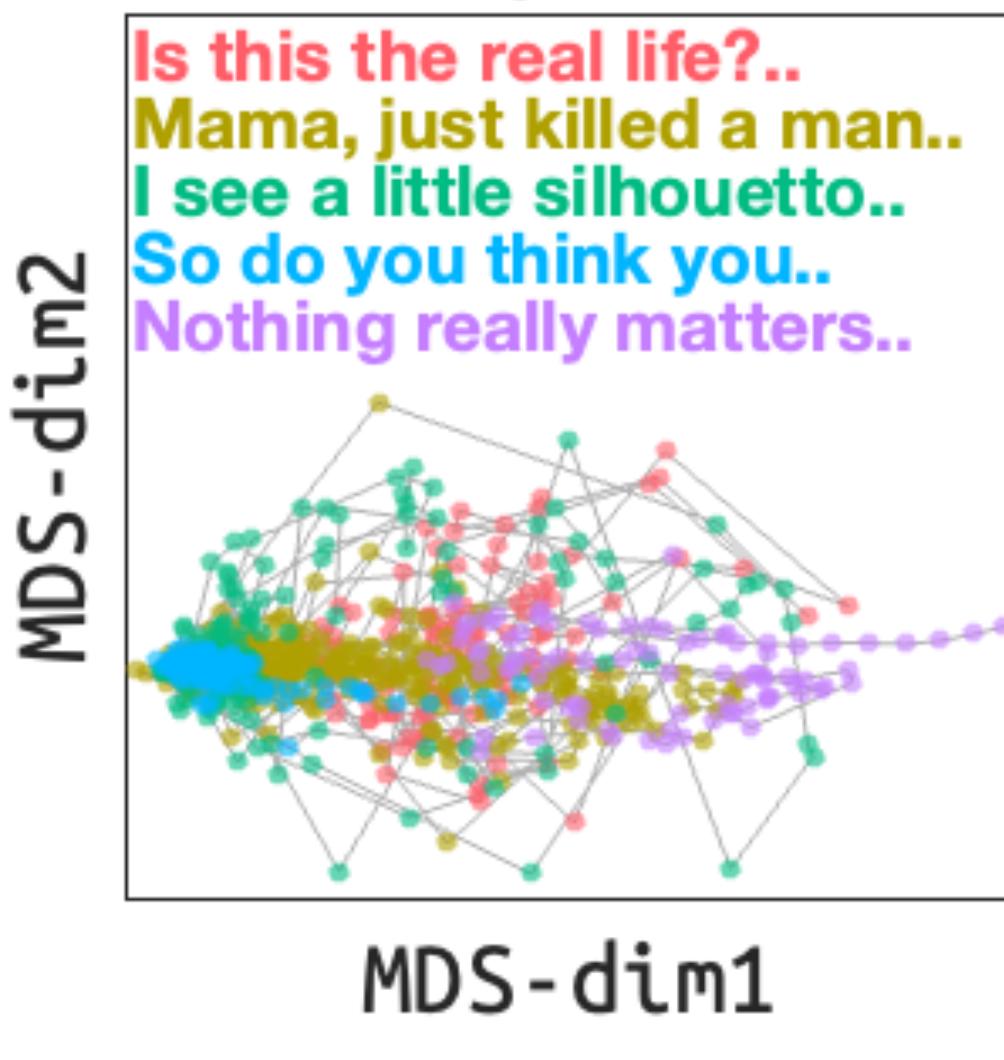
L23:EmbeddingBatch



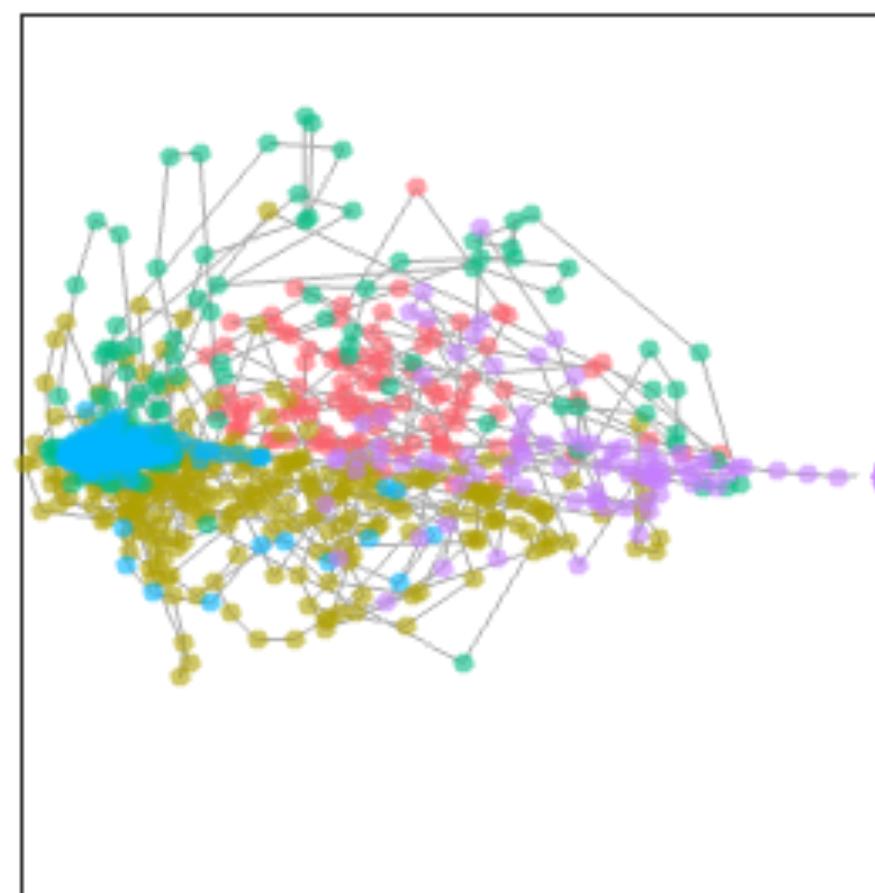
Example: Queen. (1975). Bohemian Rhapsody

VGGish representations at various layers (from 1 to 23)

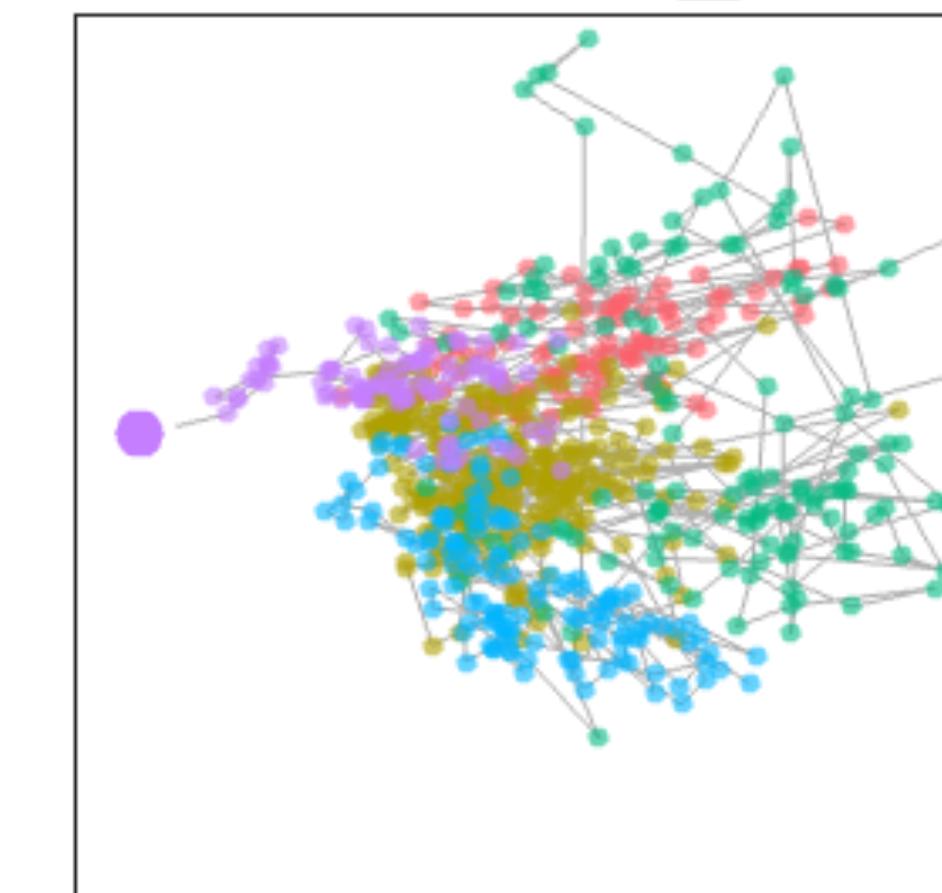
L01:InputBatch



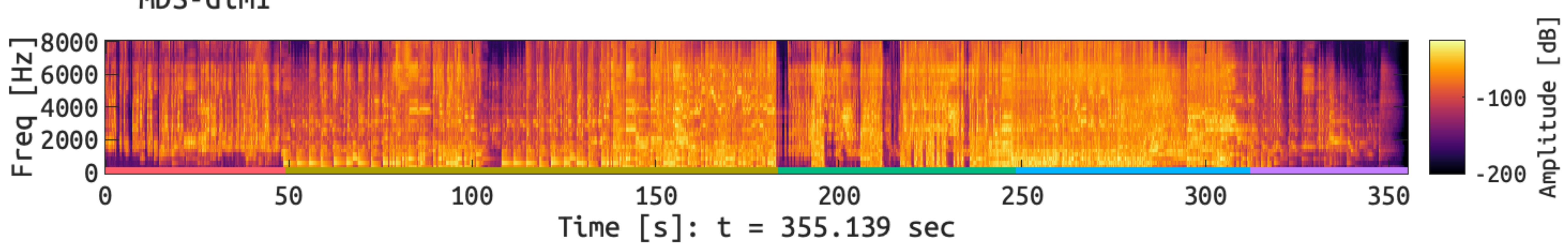
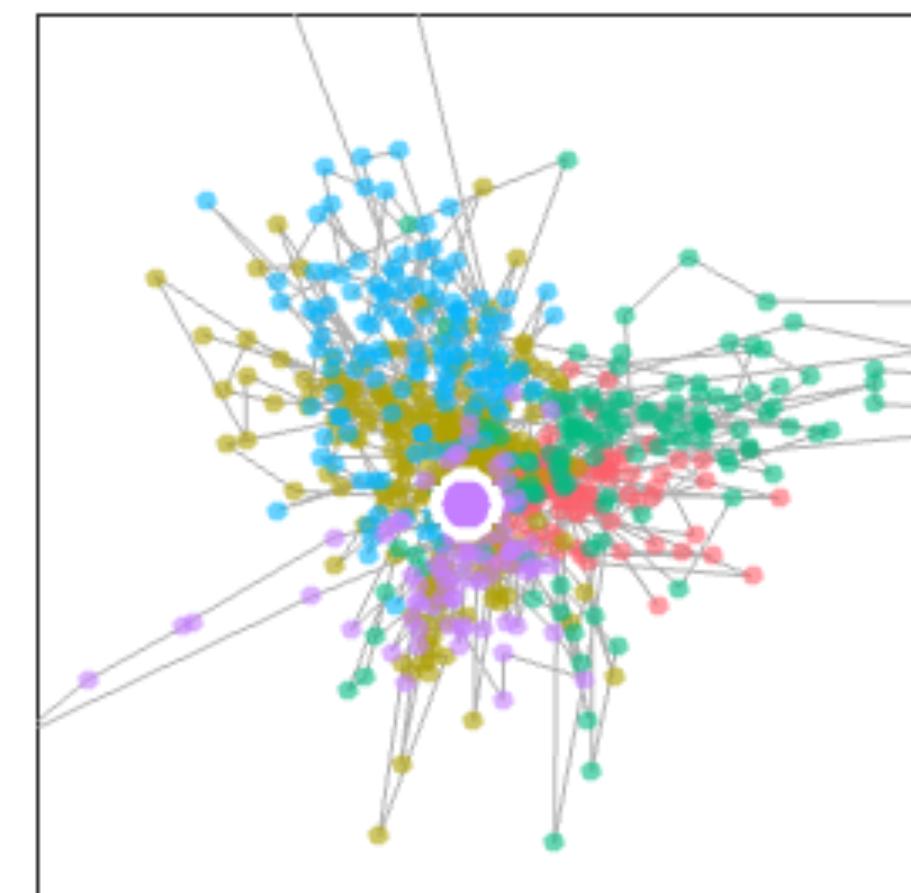
L02:conv1



L18:fc1_1



L23:EmbeddingBatch



CNN embedding for music emotion *recognition*

Potentially mid/high-level representation of music signal

An **audio semantic model called VGGish**
is based on MFCC

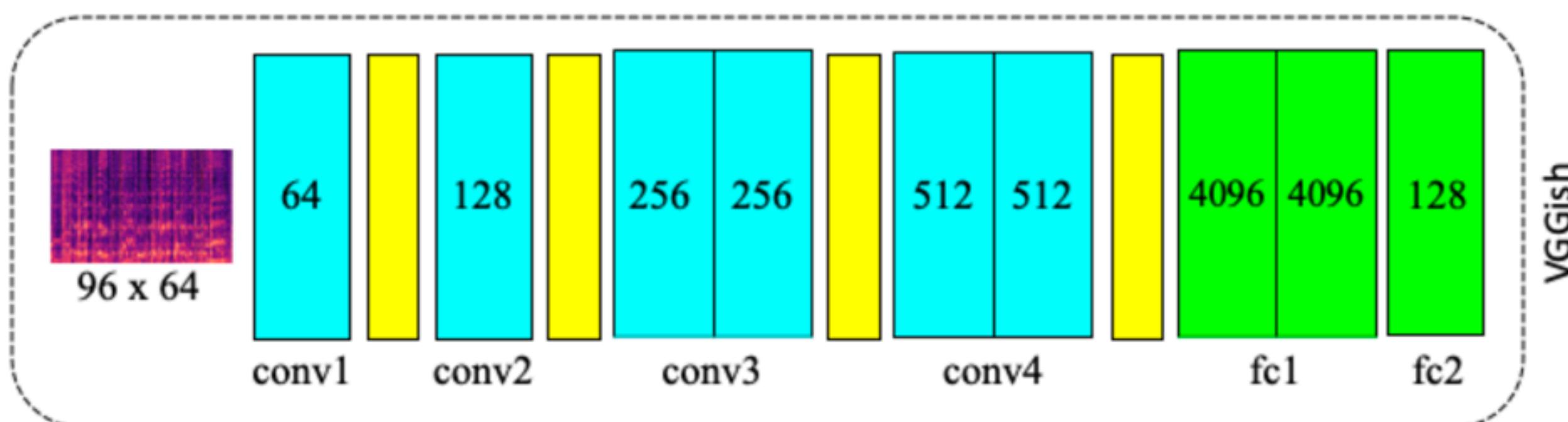
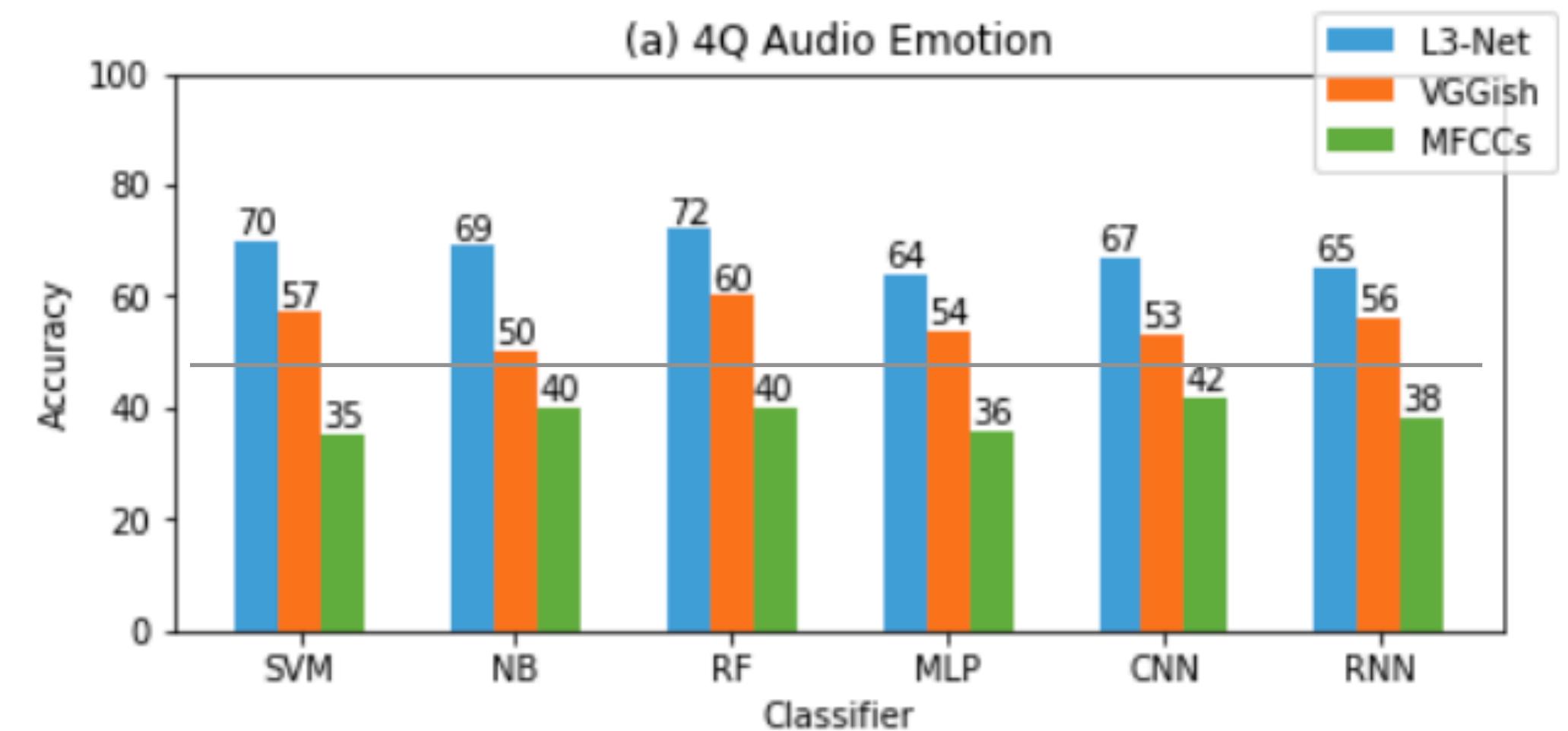


Diagram from Koh & Dubnov, 2021, ACA

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



Koh & Dubnov, 2021, ACA

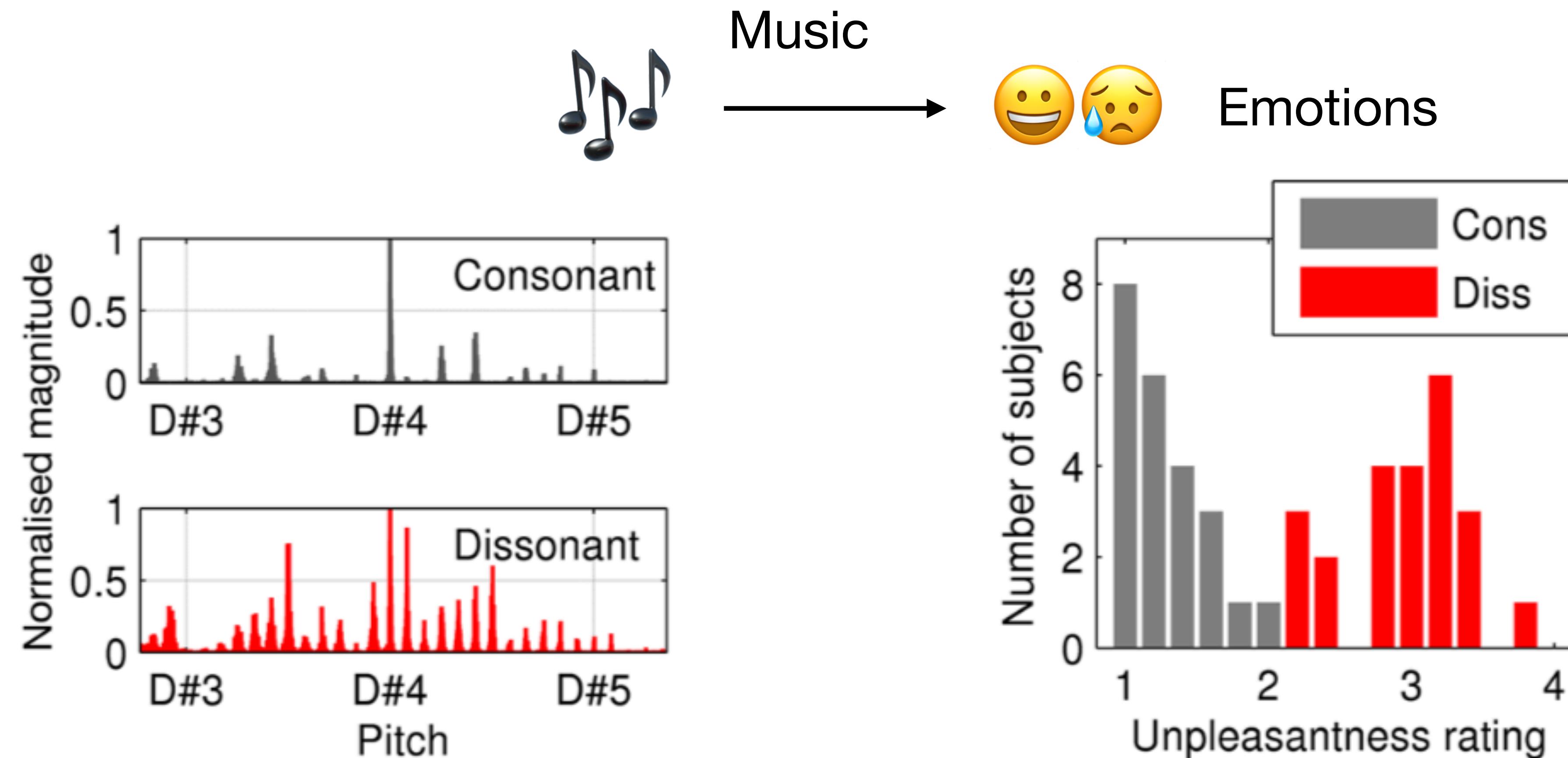
Deep audio semantic models carry more information related to expressed emotions than a traditional audio descriptor.

Topics: Introduction

Music, Brain, Emotions

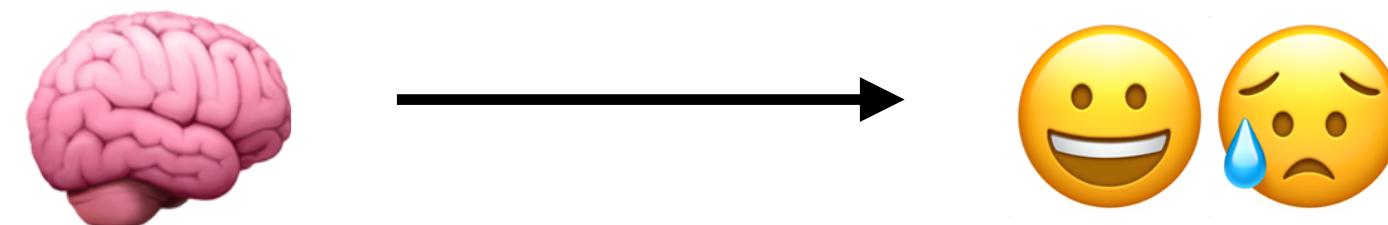
- Psychological models of emotions and musical emotions 🎵
- Cognitive accounts: Musical structures 🤔
- Affective accounts: Musical expressions 😊
- Needs for “naturalistic” experiments 💁

E.g., Consonant vs. Dissonant music

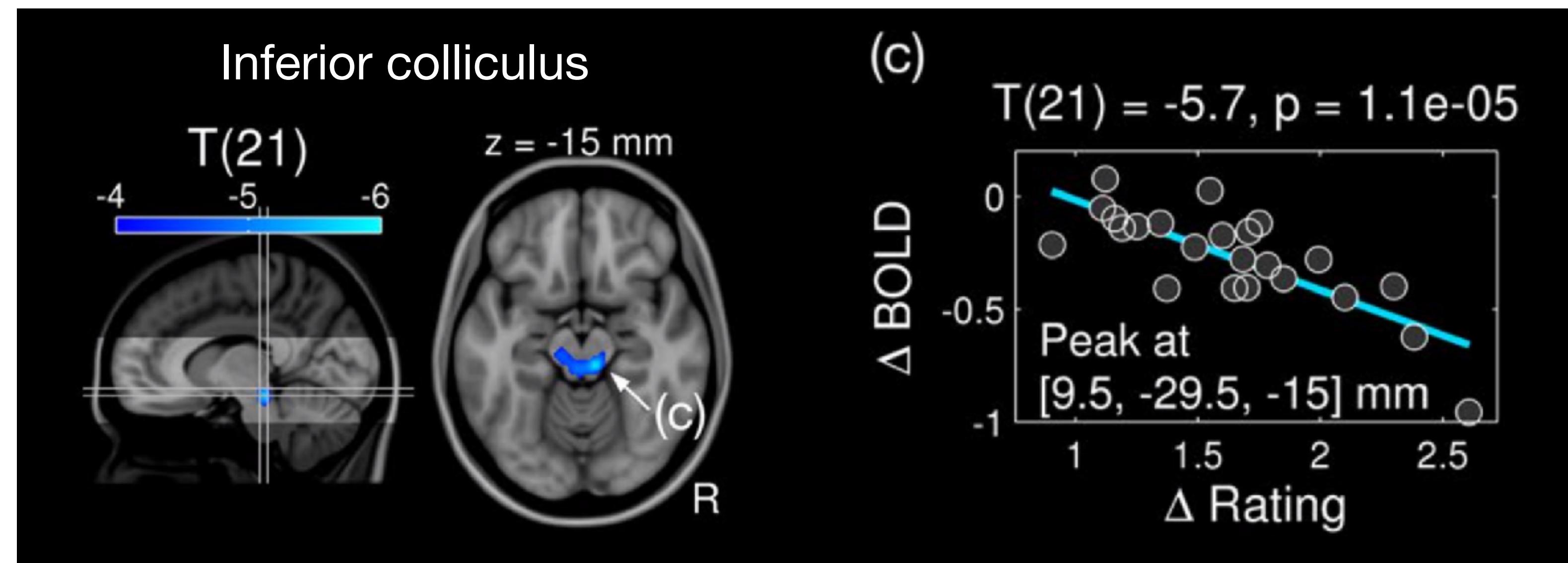


“Dissonant” stimuli were rated more unpleasant.

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

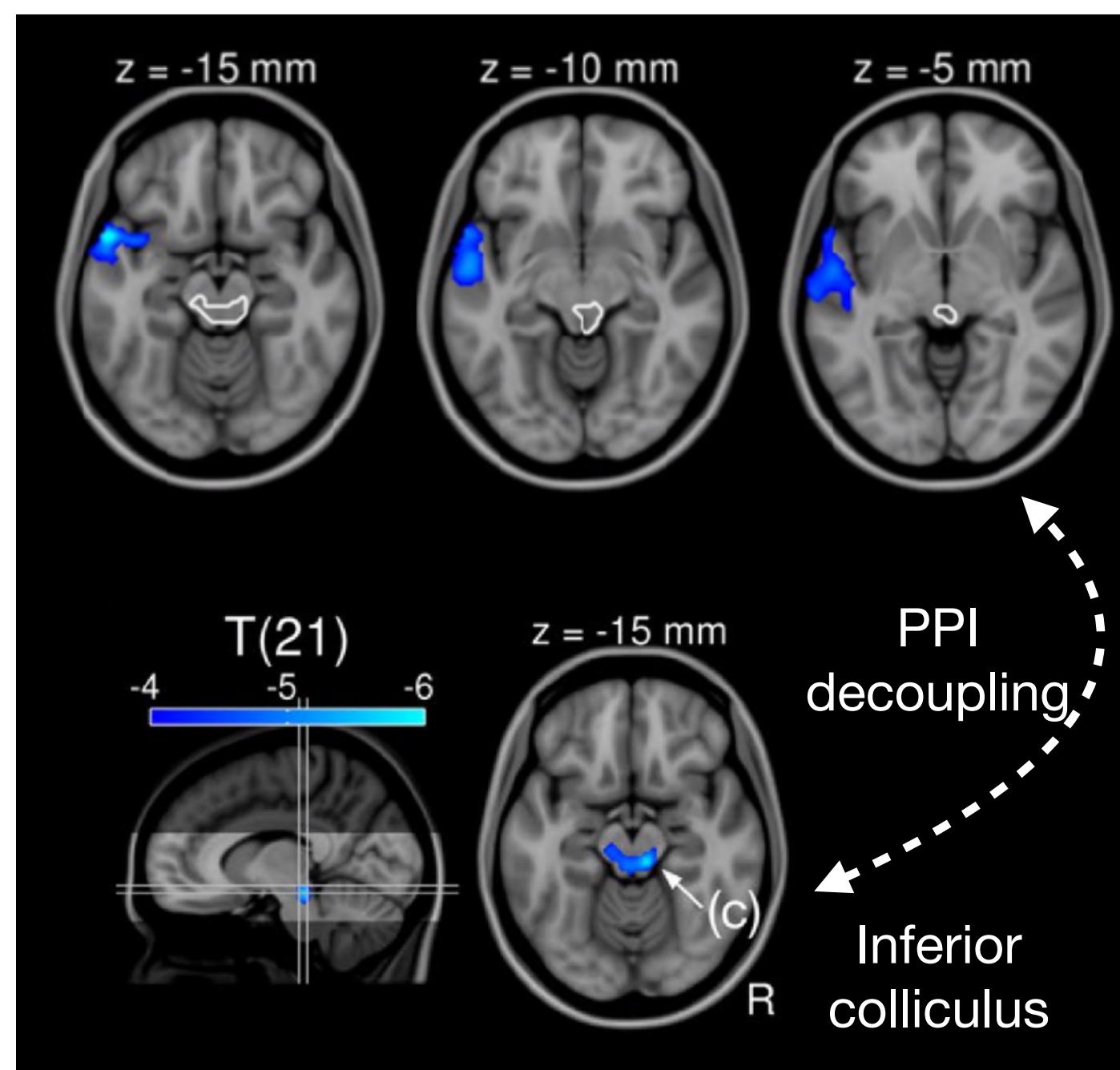


$$\Delta BOLD = \beta_0 + \beta_1 \Delta Rating + error$$

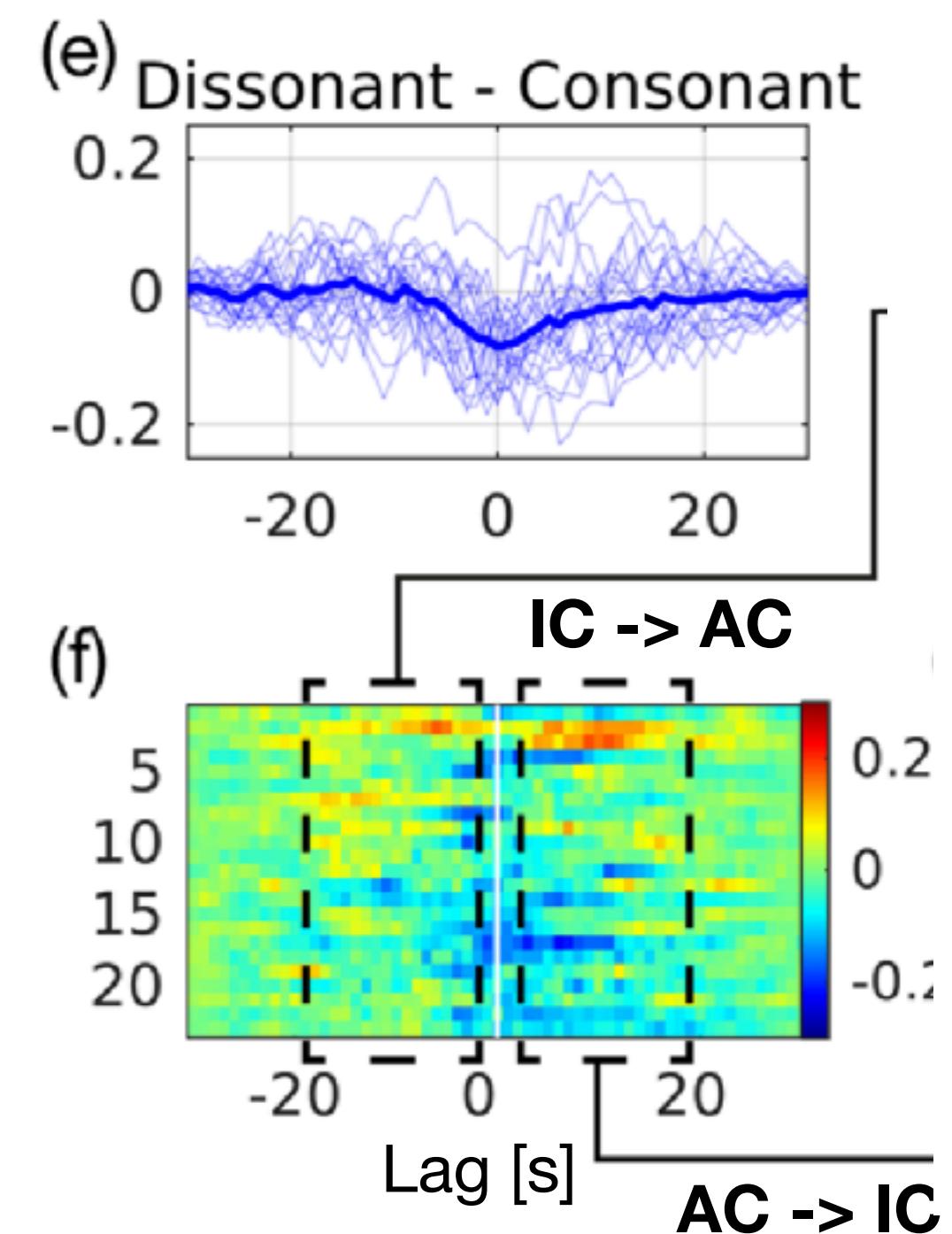


BOLD activation in the inferior colliculus (IC) decreased more in individuals who rated dissonant versions worse.

Psycho-Physiological Interaction (PPI)



Cross-correlation

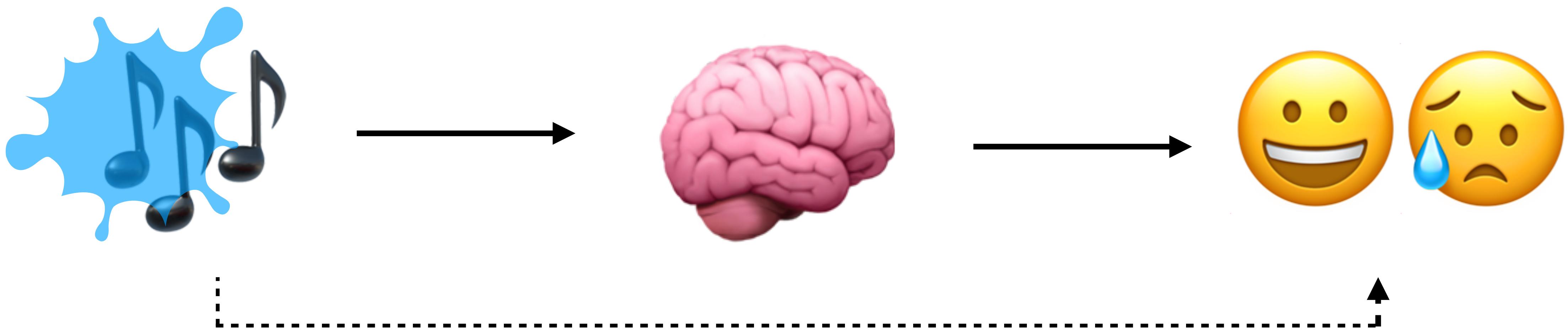


The IC decoupled more from the left auditory cortex (top-down) in individuals who rated dissonant versions worse.

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.

Summary: Introduction

Music, Brain, Emotions

- **Musical emotions** 🎵: basic emotions, dimensions, and aesthetic emotions
- **Cognitive accounts** 🤔: playing with expectation, tension & resolution,
- **Affective accounts** 😊: emotional vocalization mimicking & contagion, entrainment
- (Mnemonic accounts 💭: autobiographical memories, conditioning, imaginary, semantic associations, lyrics, ...)
- **Naturalistic experiment** 🦋: necessary complexity to investigate high-level processes

Any questions?

Introduction



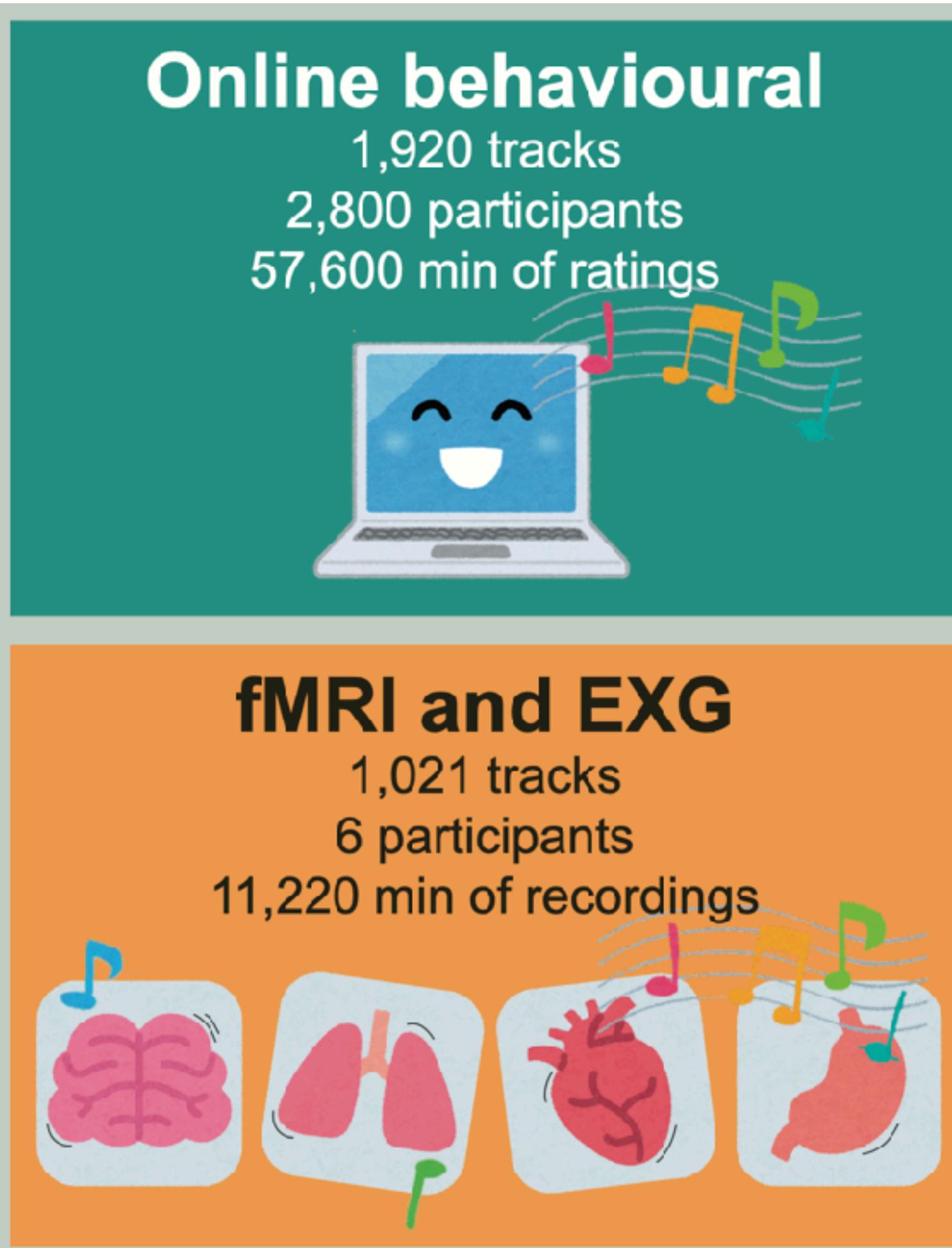
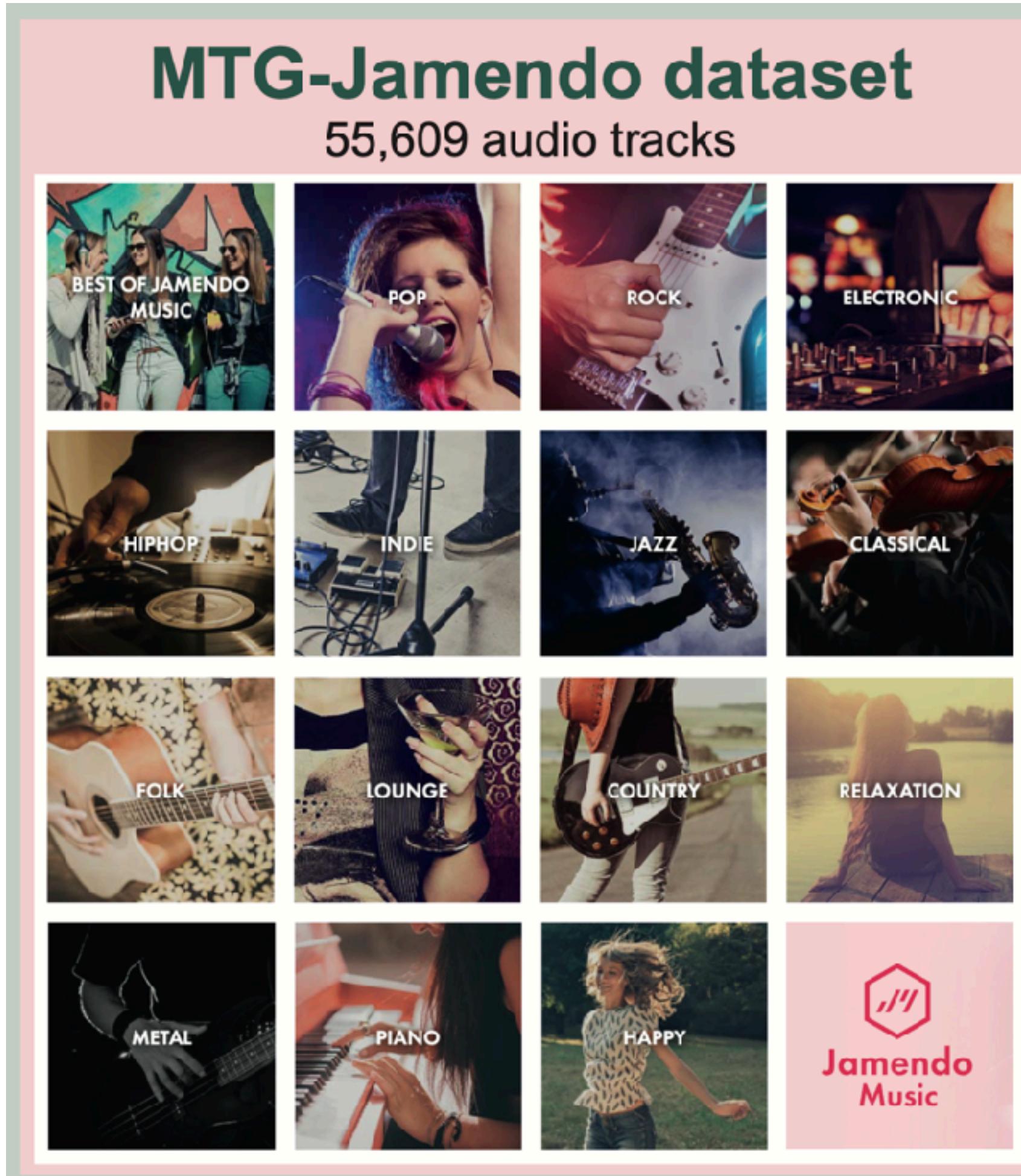
Music & Emotions

Live-experiment

[Go back to SESSION PLAN](#)

Ongoing study: "ManyMusic"

Deep phenotyping of musical emotions

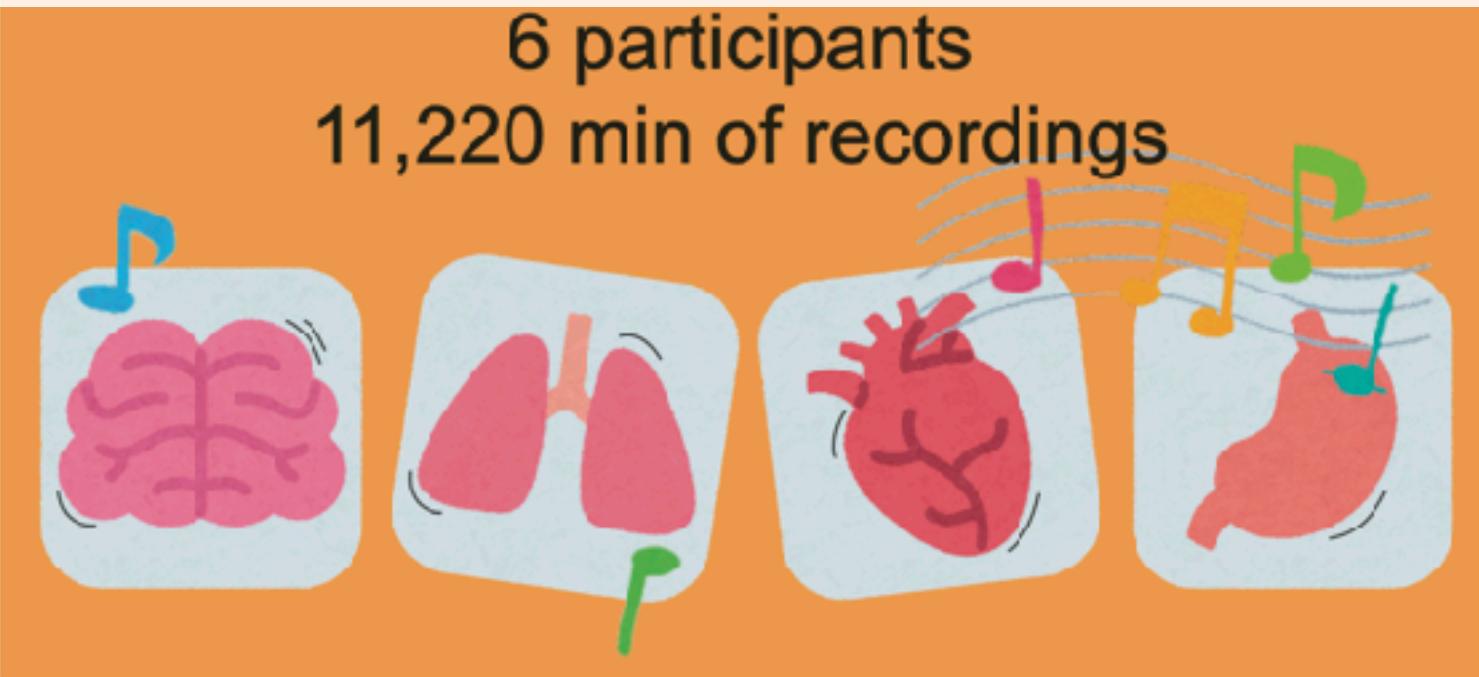
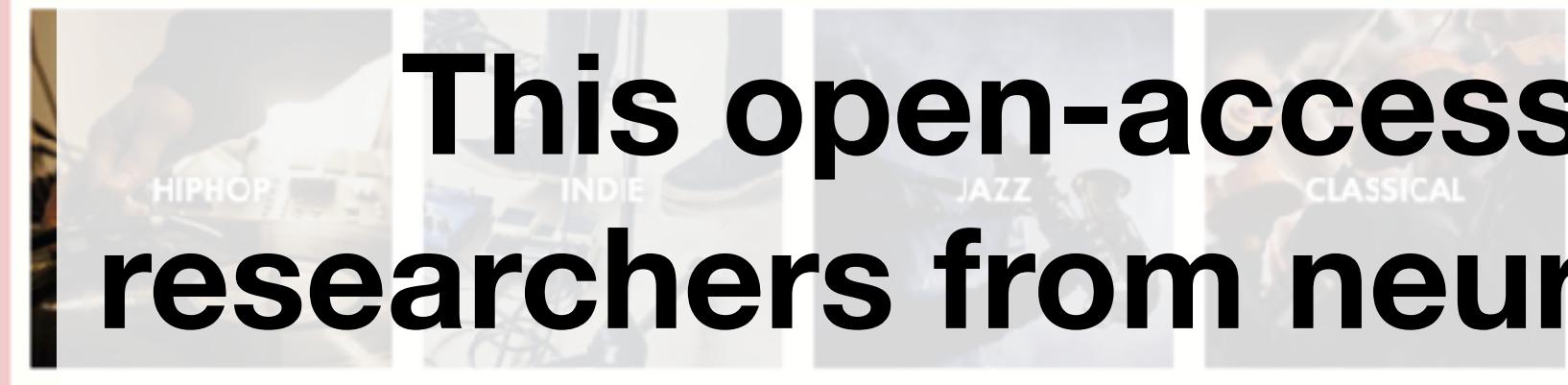
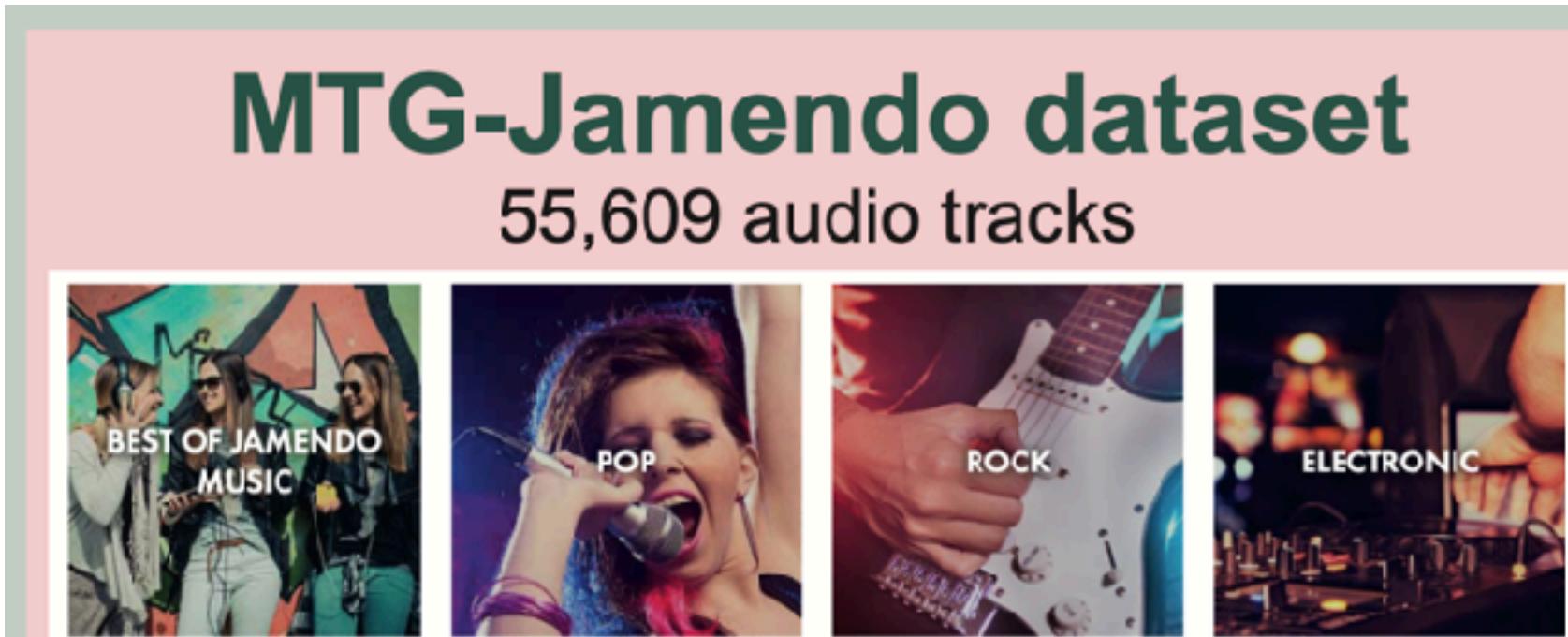


- **Stimuli:** N = 1,920
 - High-quality, diverse in styles and emotionalities (undergoing with Music Tech Group @UPF)
- **Participants:** N = 2,800
 - Prolific English speakers with minimal (2+ yr) musical experience
- **Ratings:** N = 57,600
 - 1-min around "interesting" points based on model prediction
 - 3 different "affective planes" combining 6 scales (e.g., Arousal, Valence, Liking, ...)

Ongoing study: "ManyMusic" II

Deep phenotyping of musical emotions

JQ Young Academy
at Goethe



- **Stimuli:** N = 1,021
 - Top 50% from online: 960
 - Participant-selected: 60
 - Reference: "Bohemian R".
- **Participants:** N = 6
 - Musically experienced (in various genres)

emotionally
endure 36 2-hr sessions for
expansive (JQYA)
fMRI: 17+1 sessions
MBME at 3-T

- Listening & emotion-tracing (e.g., Arousal/Valence-plane)
- **EXG:** 17+1 sessions
 - 32-ch EEG, fEMG, Resp, Pulse, SCR, bipolar EGG
 - Listening & emotion-tracing

Overview of today's experiment

Pilot study of ManyMusic♪♪

- **What:** Emotional rating (self-report) while listening to various music & a few questionnaires about musical emotions
- **How long:** Would take about 20–30 minutes
- **Compensation:** Data from multiple participants! 😍
- **Now you need:**
 - a sufficiently charged laptop connected to the internet 
 - headphones or earphones 
 - a mouse, a trackpad, or any pointer device (like a gamepad, a stylus, ...) 

Study link

<https://onlineexperiment.ae.mpg.de/publix/eI XedE8yqVL>



onlineexperiment.ae.mpg.de/

Music Experiment for MPSCog

This experiment is developed as a teaching demonstration for Max Planck School of Cognition students attending Cognitive Academies in December 2024 in Berlin.

Hope you enjoy it!😊

2024-12-05, seung-goo.kim@ae.mpg.de

Before we begin...

Purpose. This live experiment aims to investigate self-reported emotion evoked by music listening. A pedagogical goal is to have students to experience a study investigating subjective affective experience induced by music as a participant and as a researcher for better introduction to the field of affective neuroscience of music.

Procedure. A participant will listen to various music via headphones and rate their

Cancel



Time for coffee!



[Go back to SESSION PLAN](#)



How to analyze naturalistic experiments

Theories

Topics: Analysis

How to analyze naturalistic experiment

- Data-driven methods:
 - **Intersubject synchrony**
 - **Event structure analysis**
- Hypothesis-based methods:
 - **Linearized encoding analysis**

Topics: Analysis

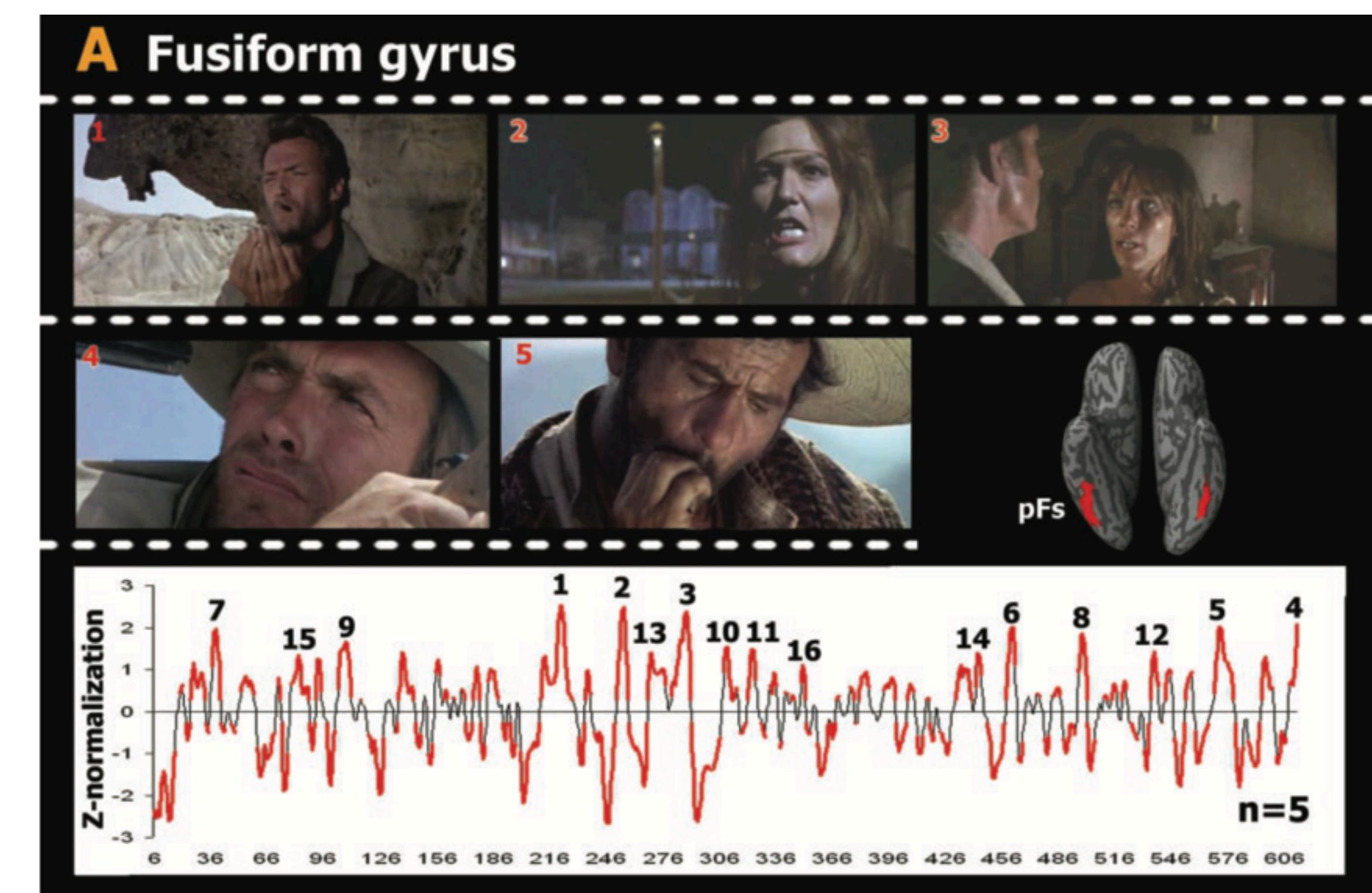
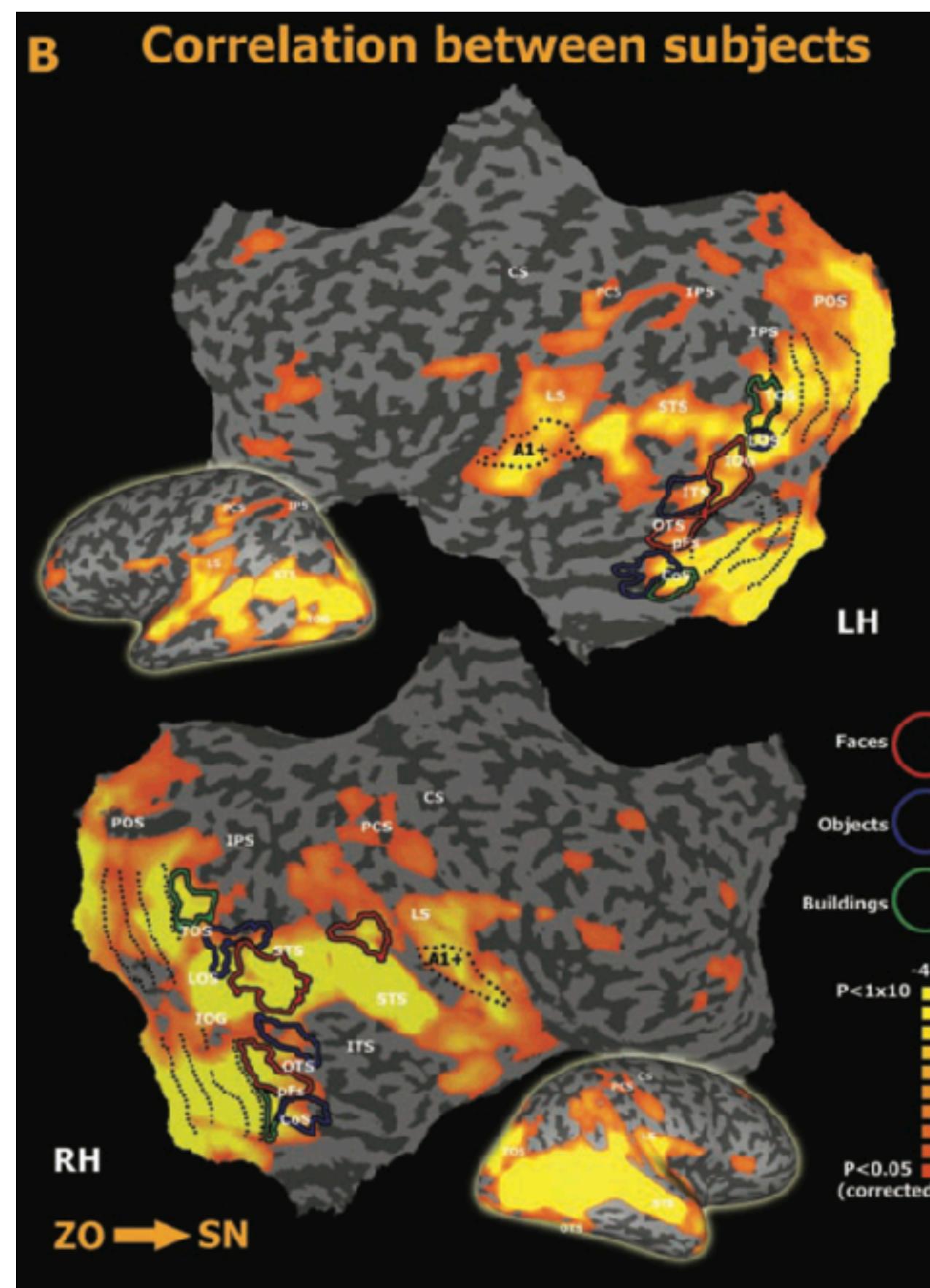
How to analyze naturalistic experiment

- Data-driven methods:
 - **Intersubject synchrony**
 - **Event structure analysis**
- Hypothesis-based methods:
 - **Linearized encoding analysis**

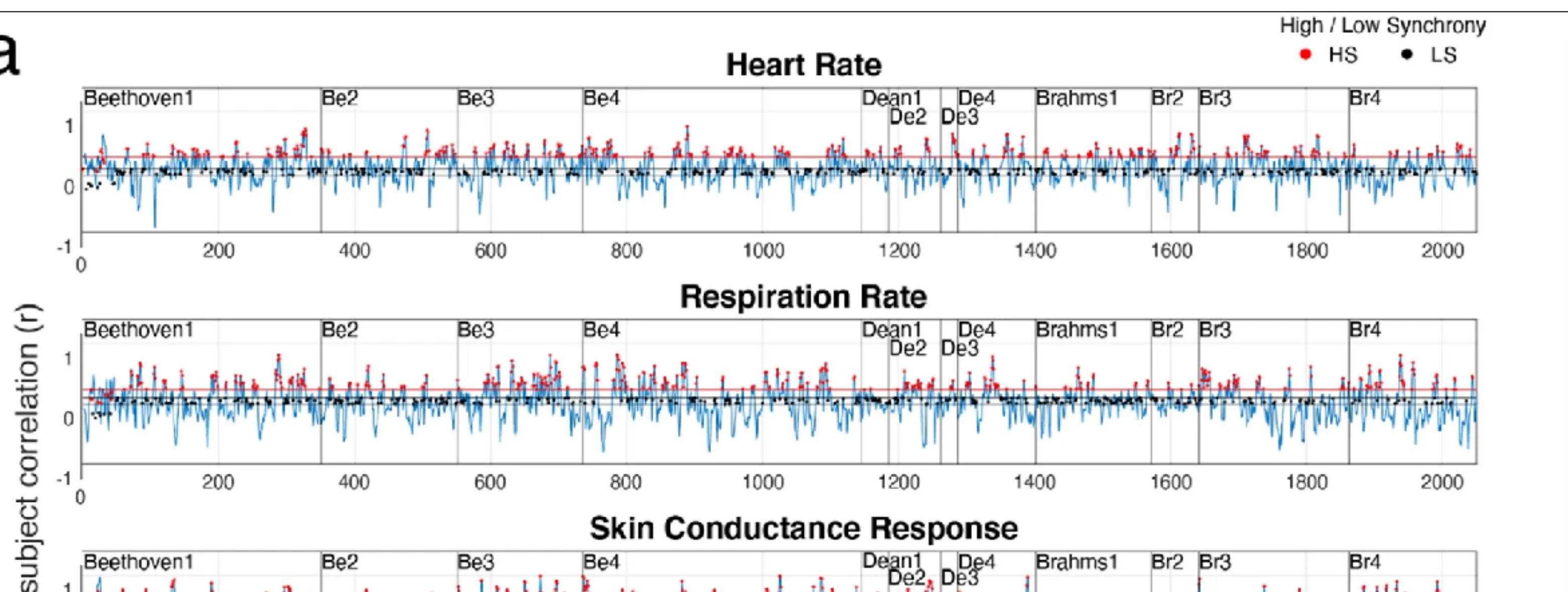
Intersubject synchrony during a movie

Hasson et al., 2004, *Science*

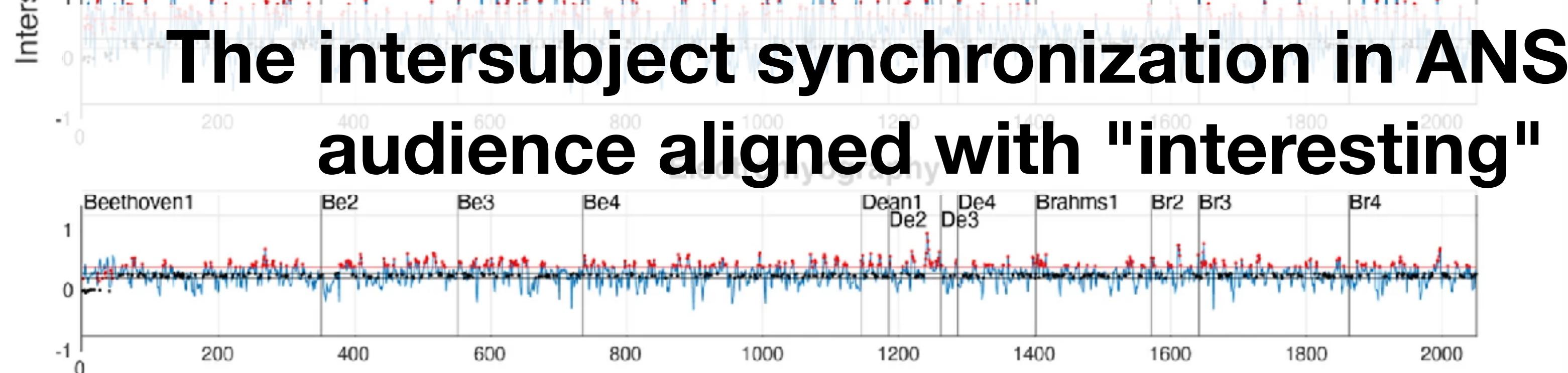
- 30-min movie
- 5 participants
- Free viewing + reporting the plot



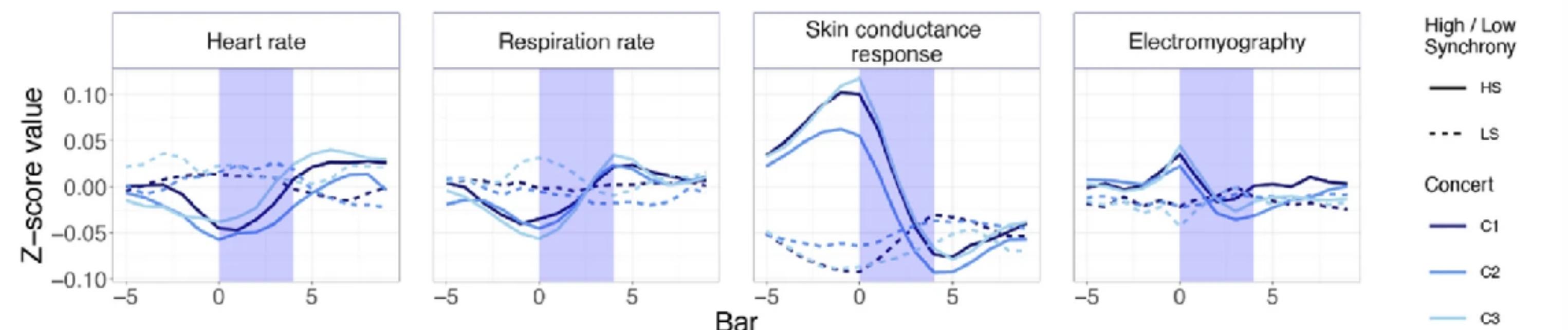
a



The intersubject synchronization in ANS measures amongst audience aligned with "interesting" musical events.



b



Tschacher et al., 2024, *Sci Rep.*
(Experimental Concerts in 2022, Berlin)

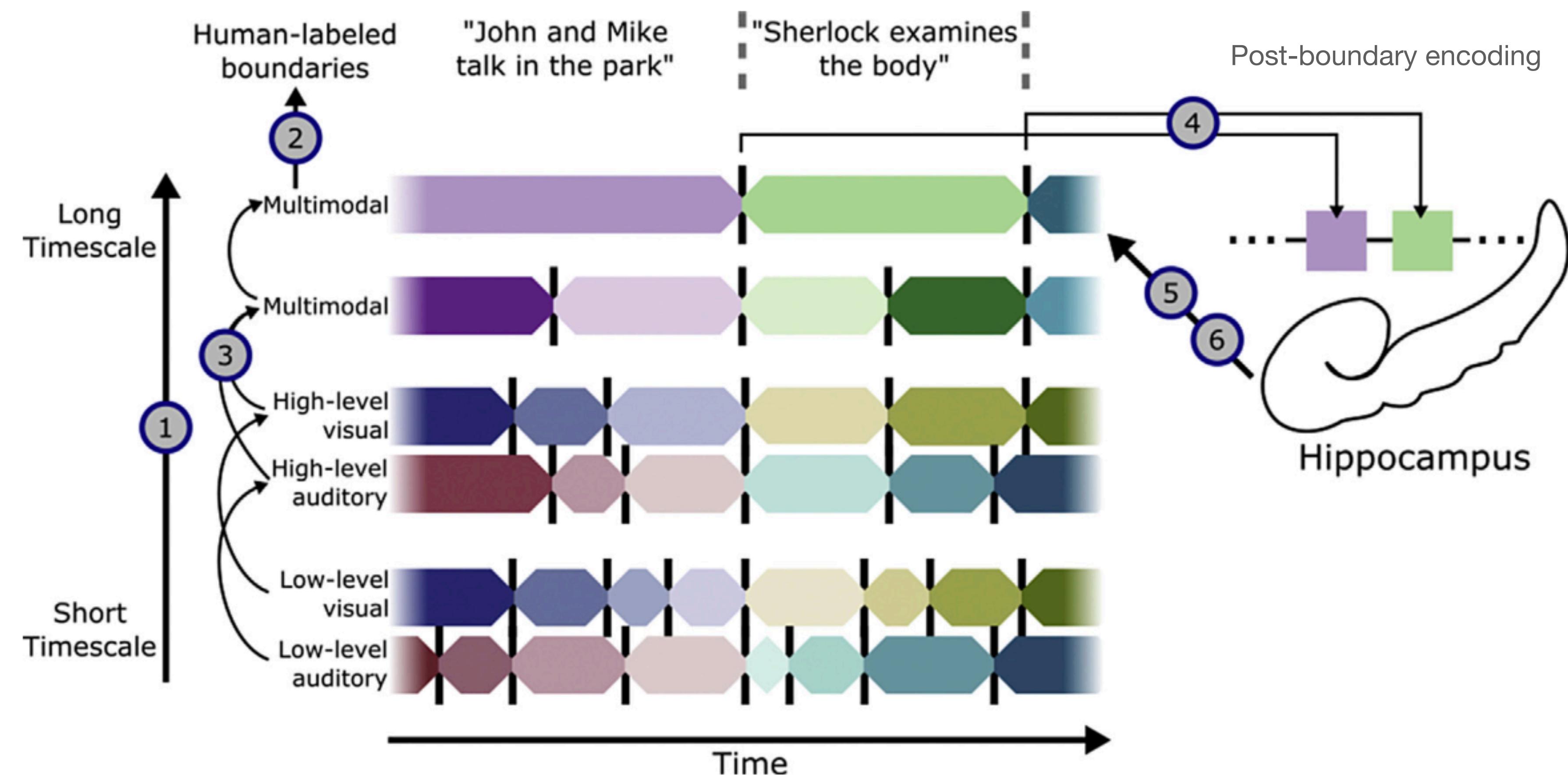
Topics: Analysis

How to analyze naturalistic experiment

- Data-driven methods:
 - **Intersubject synchrony**
 - **Event structure analysis**
- Hypothesis-based methods:
 - **Linearized encoding analysis**

Theory of "event segmentation" & memory

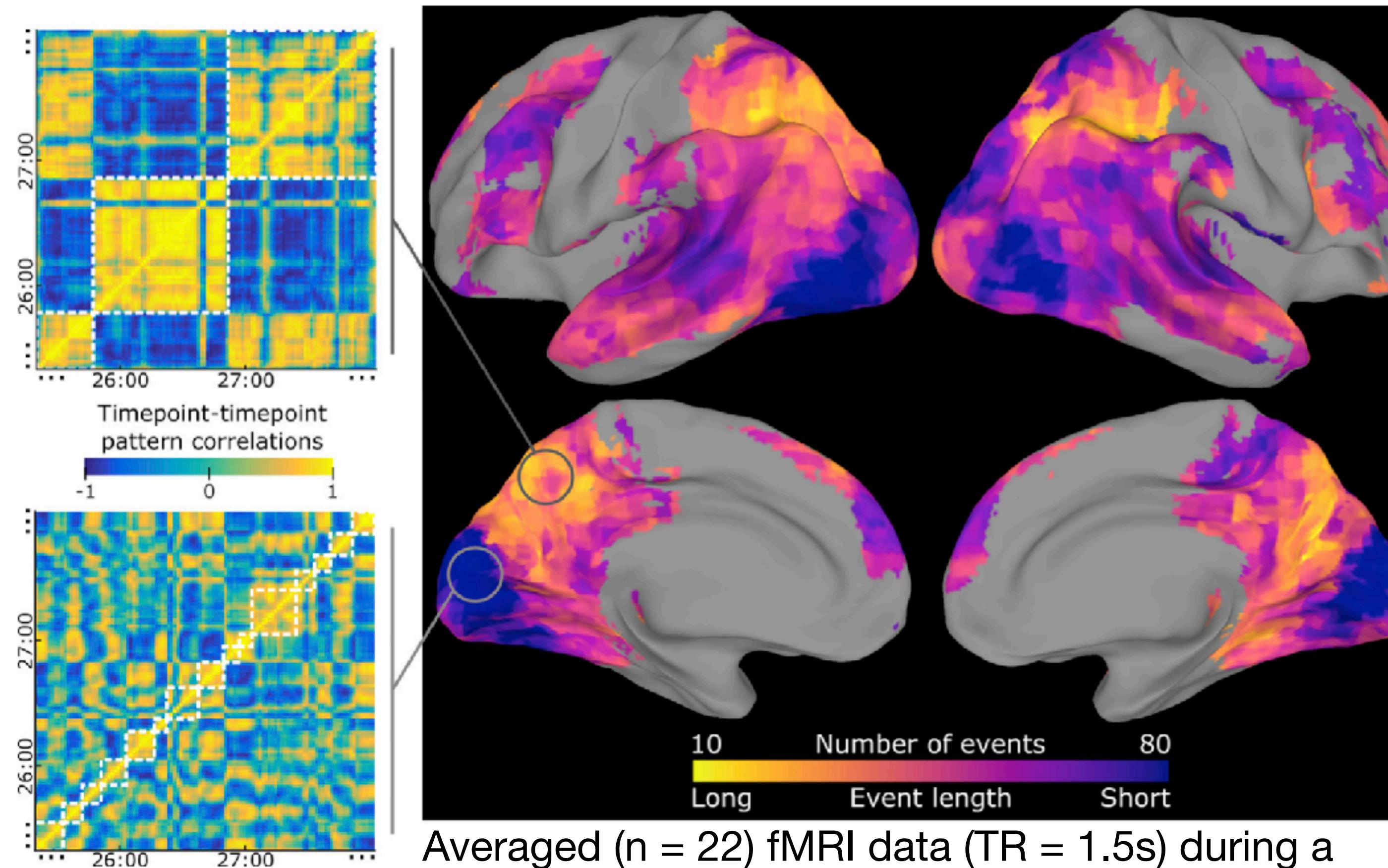
Baldassano et al., 2017, *Neuron*



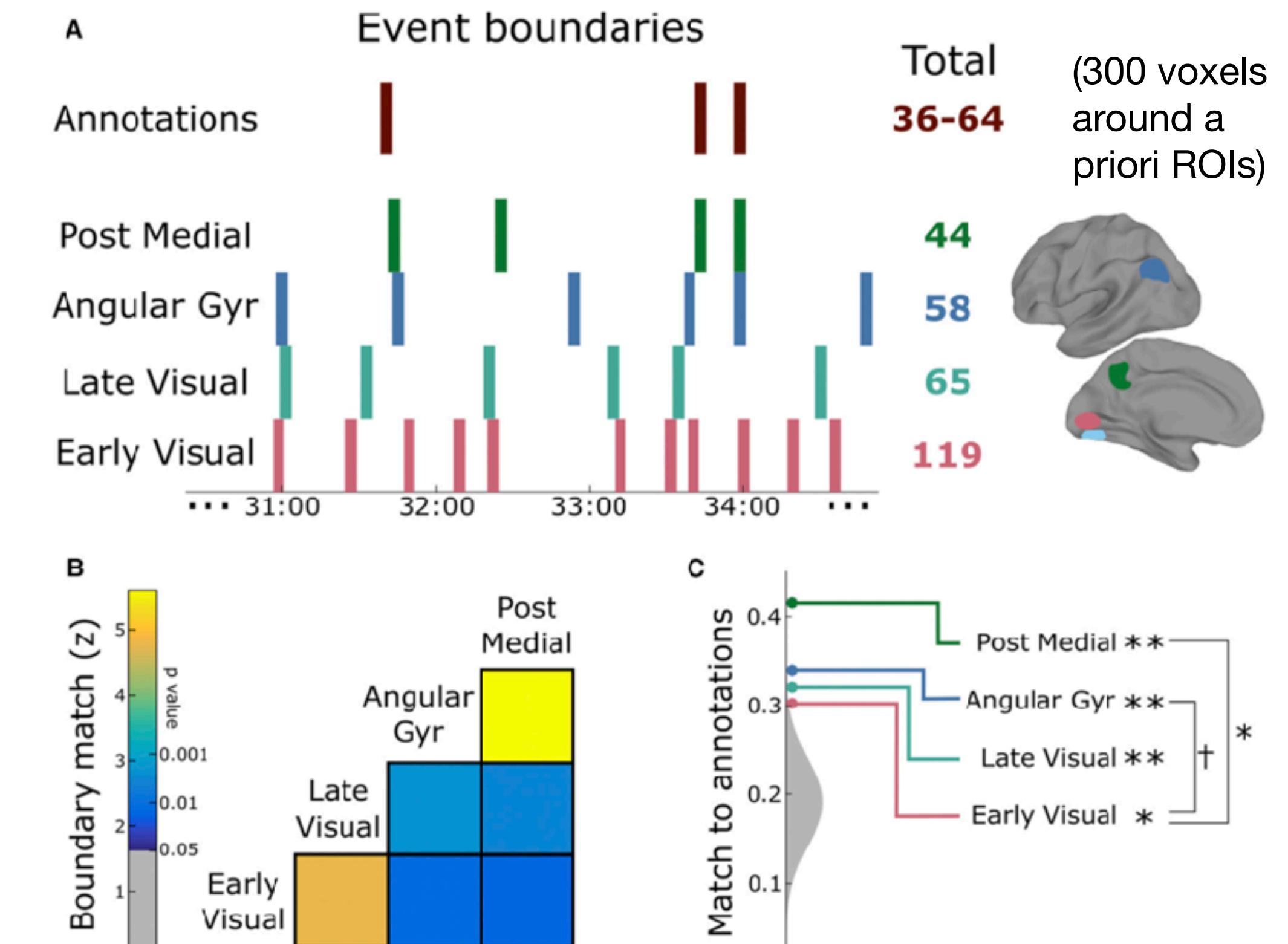
SEE ALSO Zack et al., 2007, *Psychological Bulletin*, DOI:10.1037/0033-2909.133.2.273

Event time scales during a movie

Baldassano et al., 2017, *Neuron*



($n=4$; "major shifts in location, topic, time")

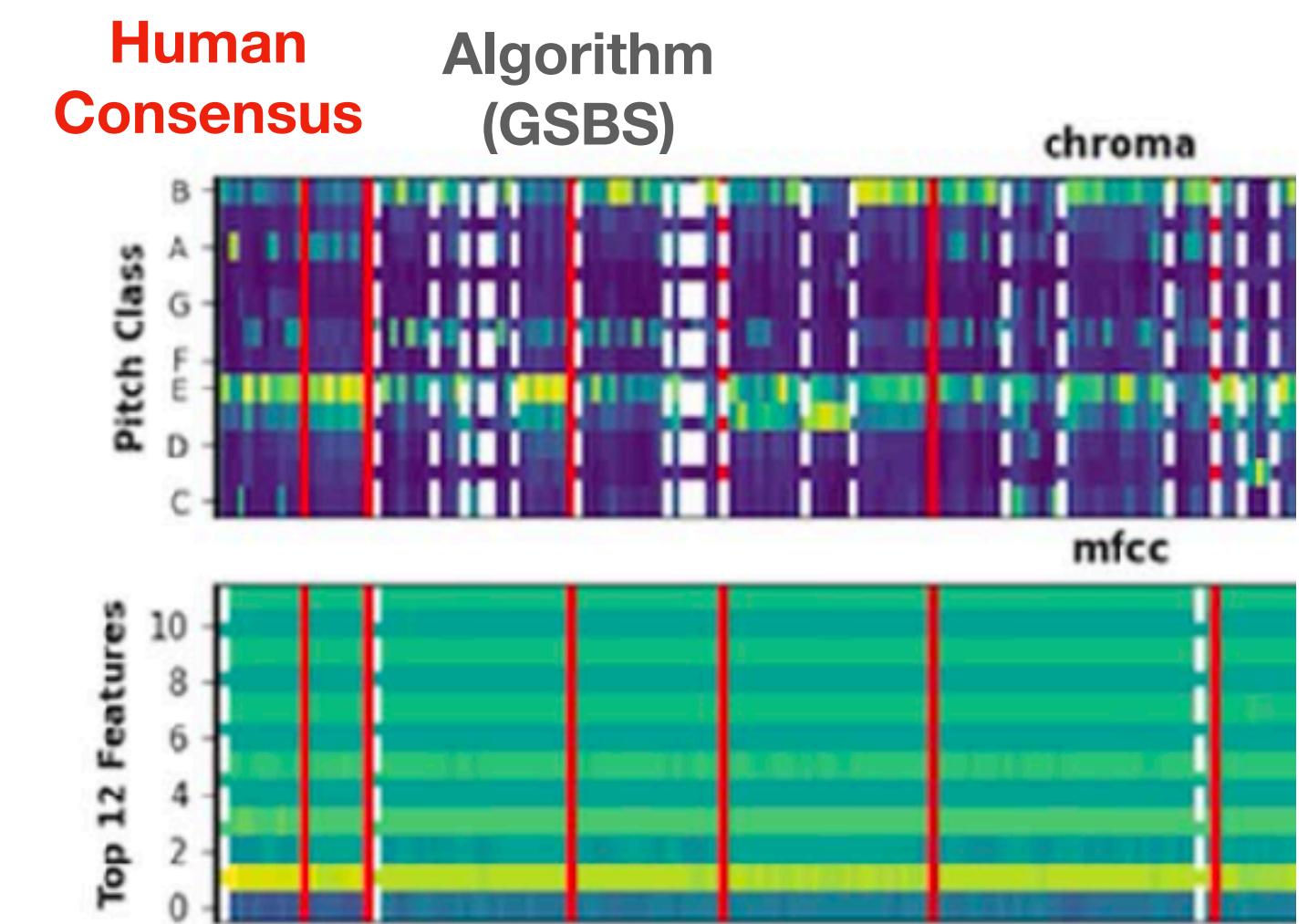
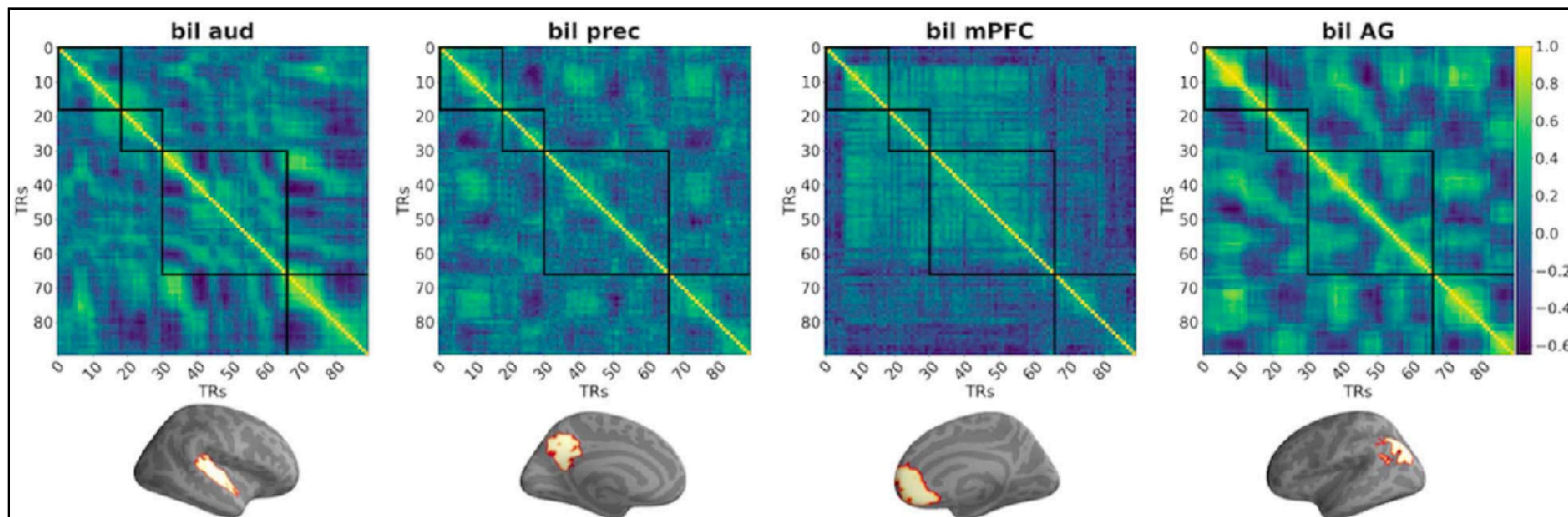
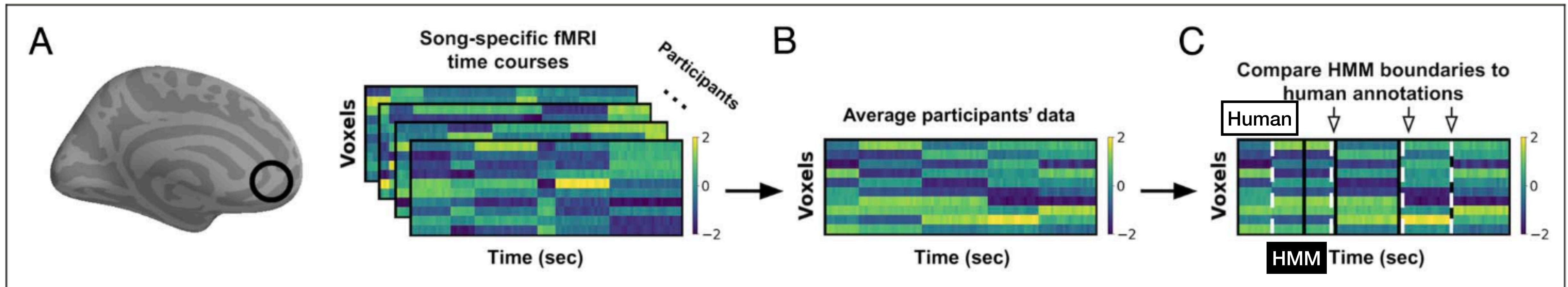


Comparison with human annotations vs. fMRI boundaries



Event structure of music in the DMN areas

Williams et al., 2022, *J Cog Neurosci*



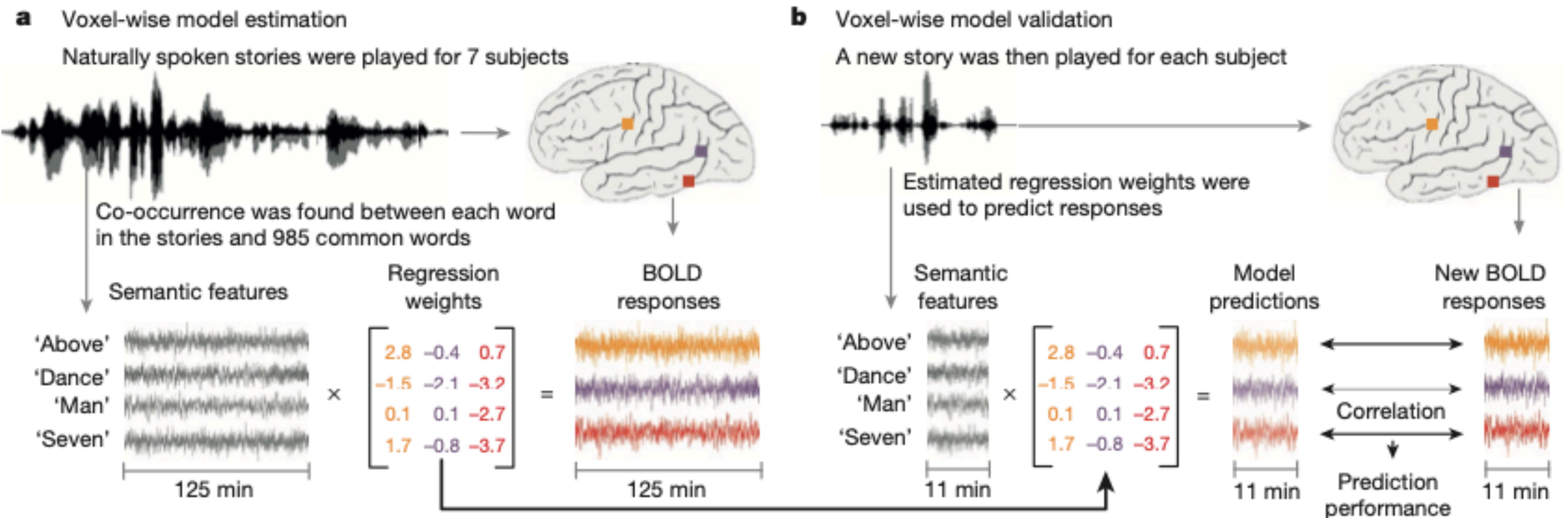
Topics: Analysis

How to analyze naturalistic experiment

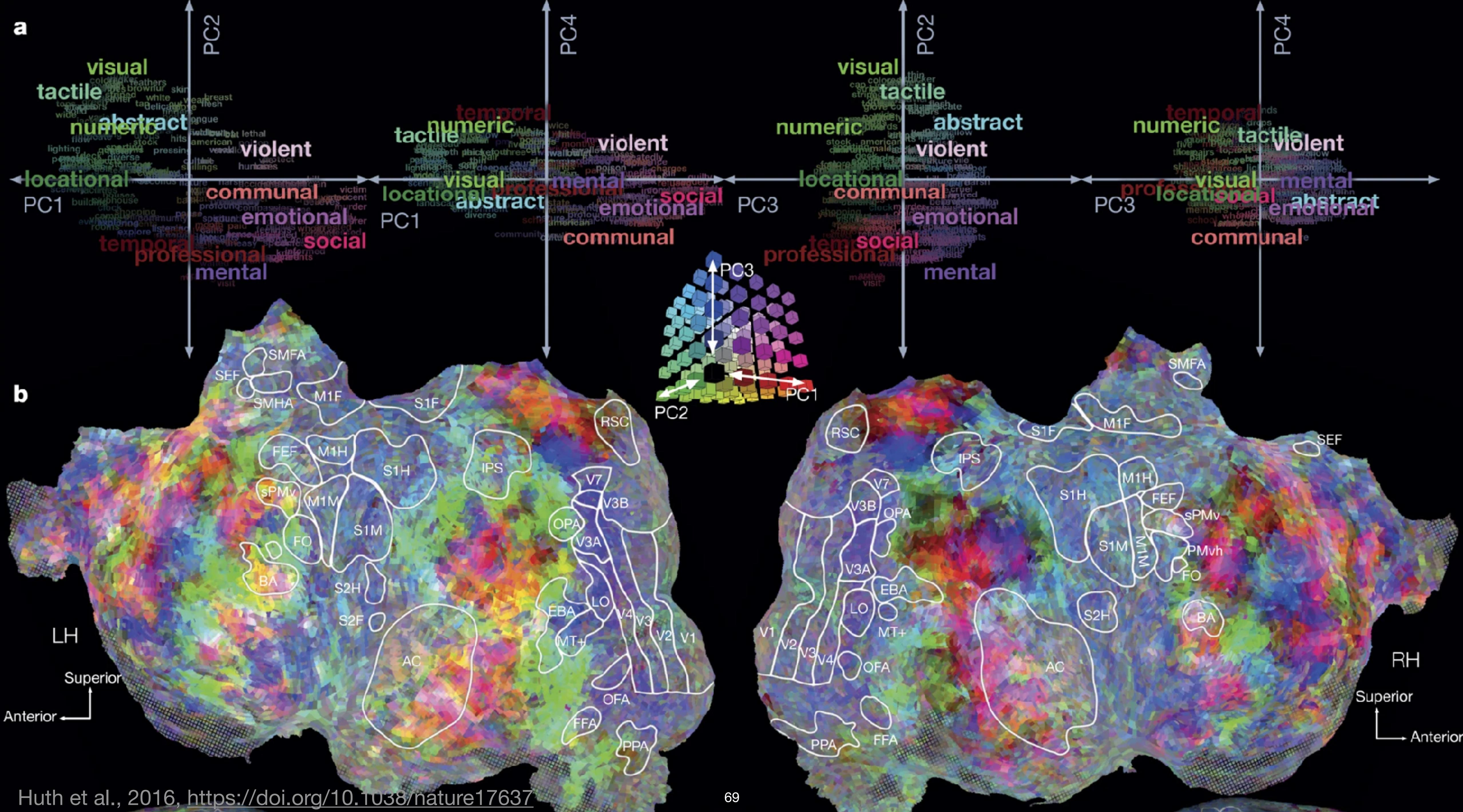
- Data-driven methods:
 - **Intersubject synchrony**
 - **Event structure analysis**
- Hypothesis-based methods:
 - **Linearized encoding analysis**

Semantic representation in the whole cortex

Non-invasive scanning of functional MRI (local paramagnetism)

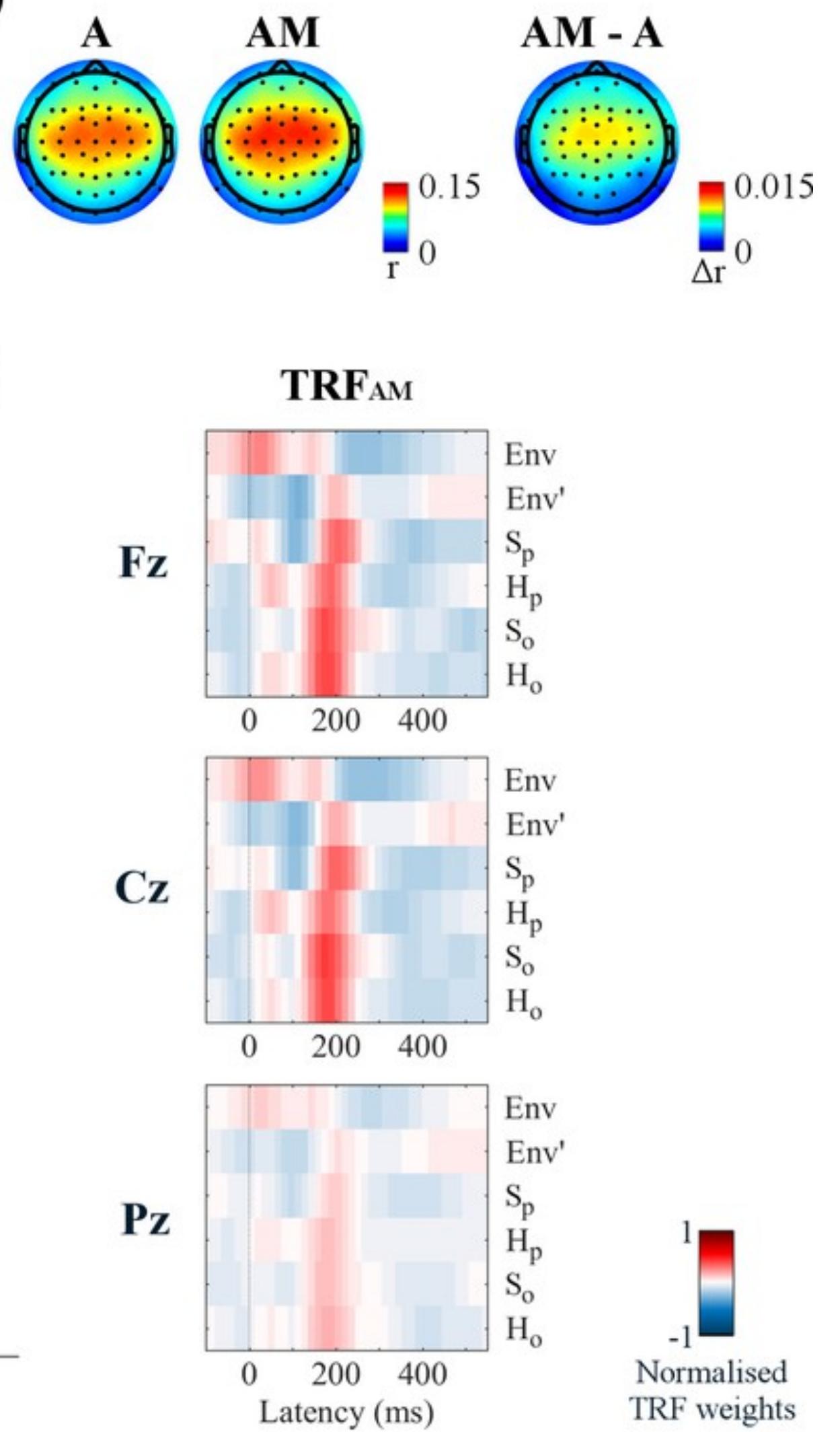
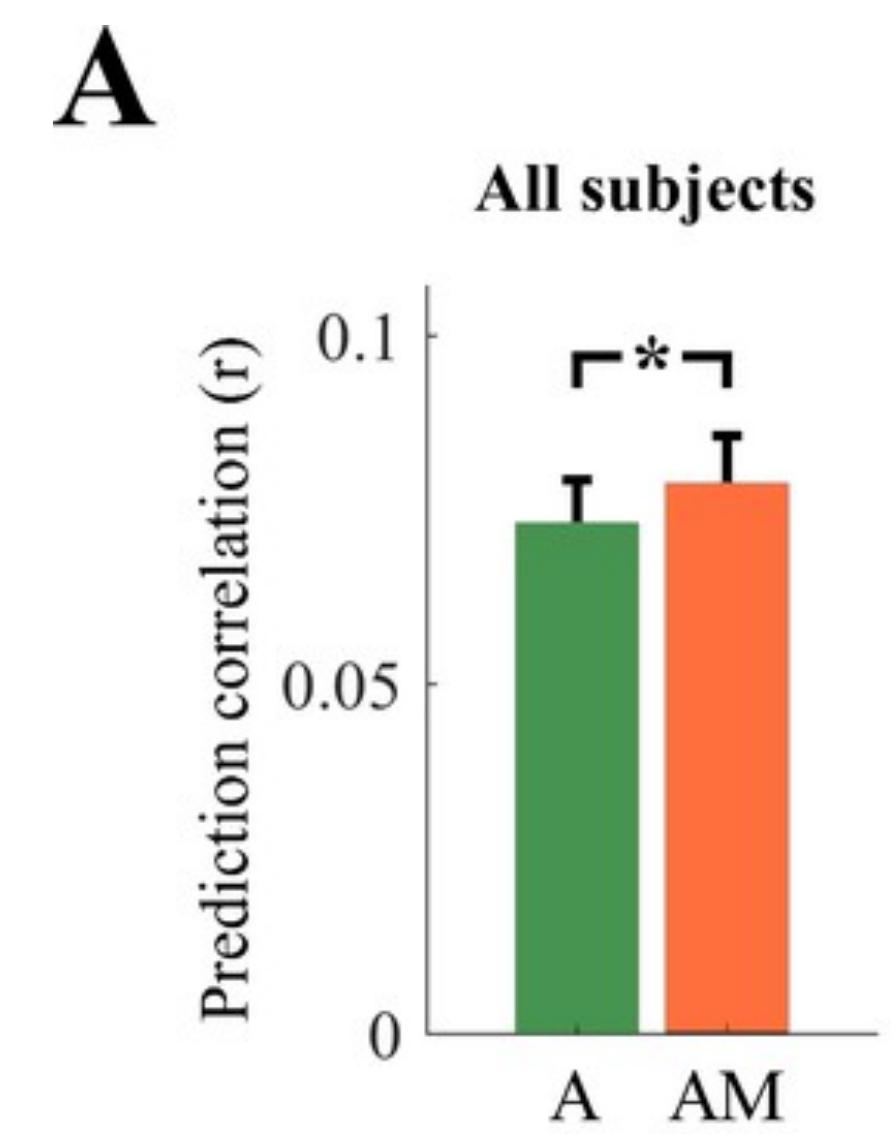
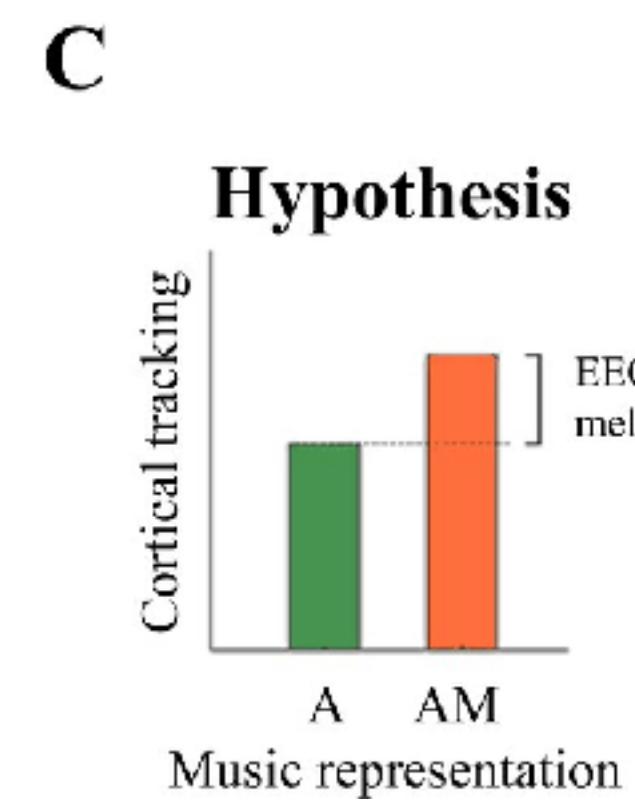
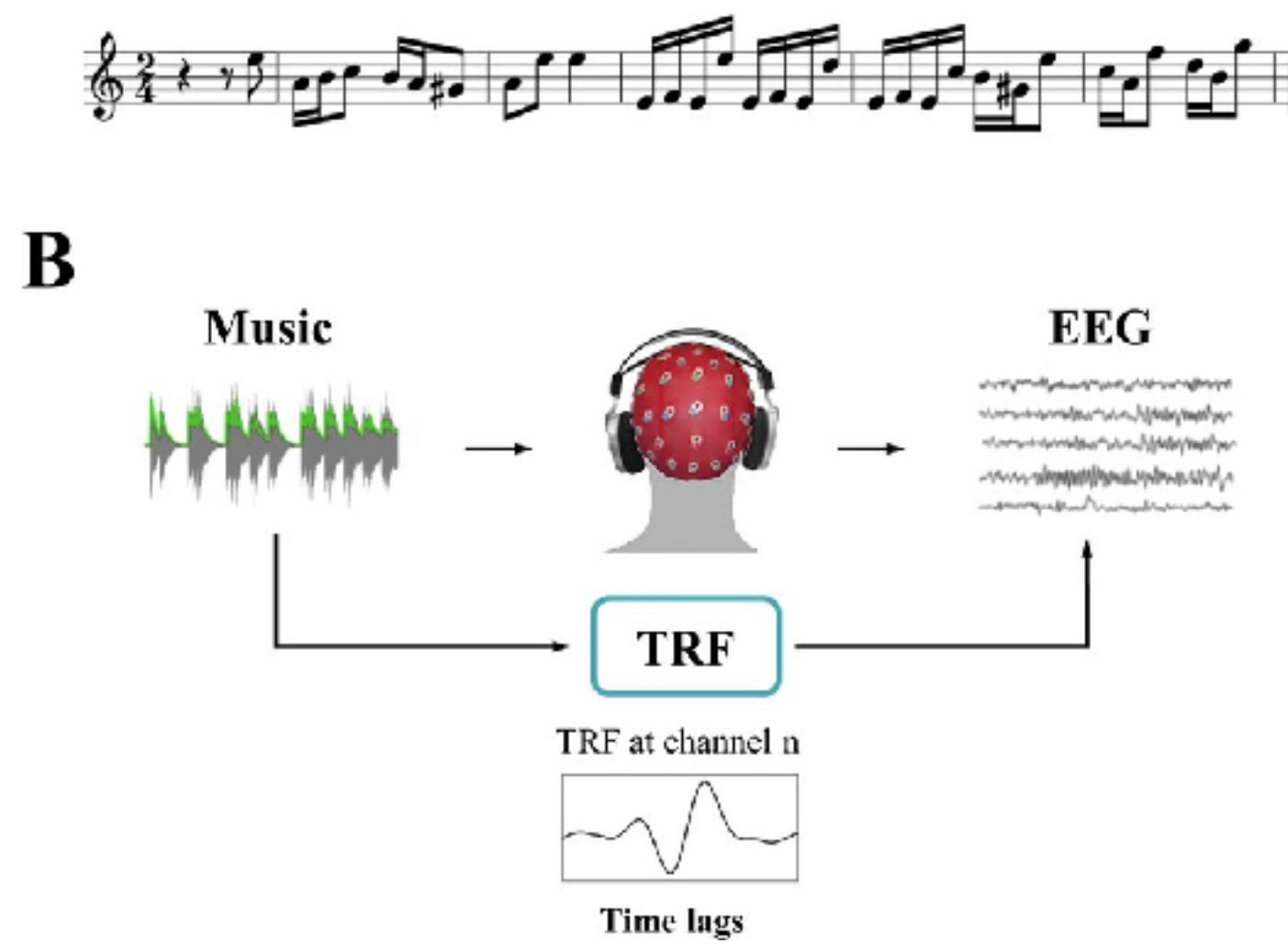
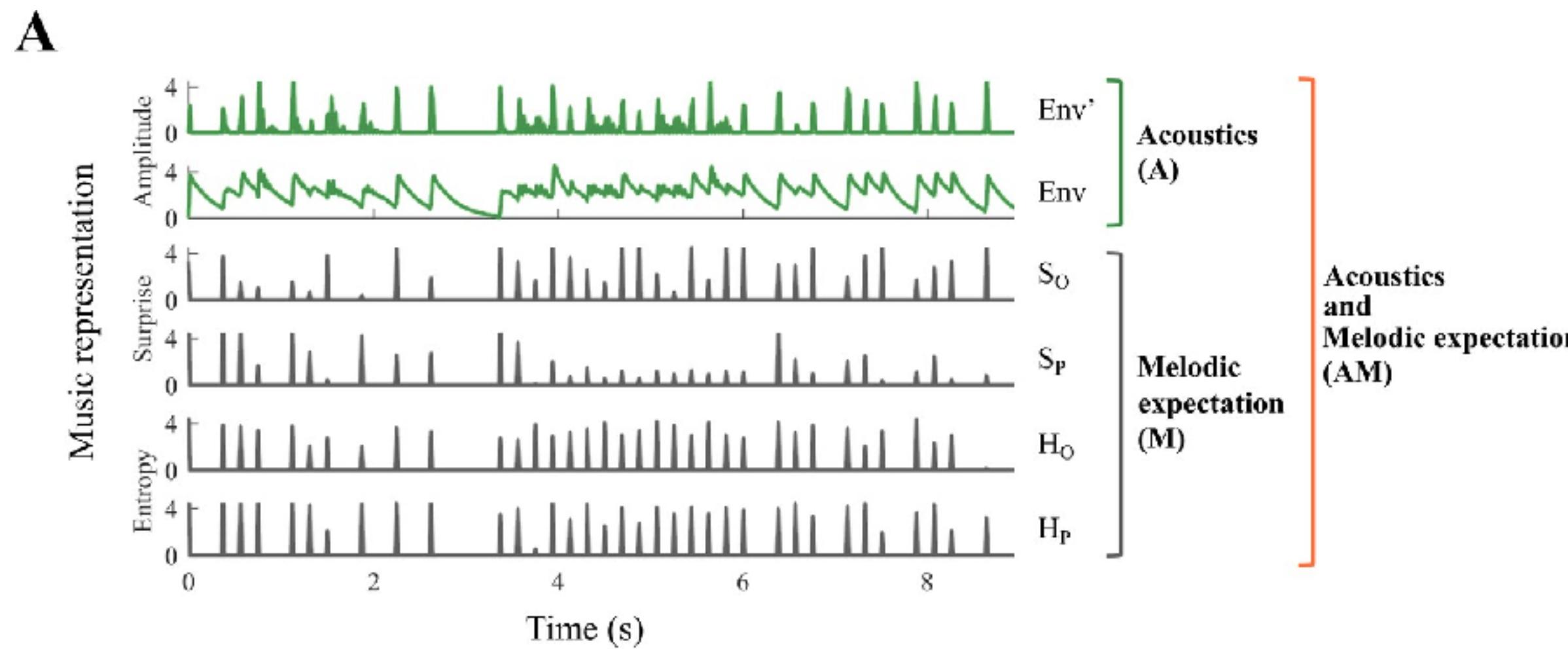


Huth et al., 2016, <https://doi.org/10.1038/nature17637>



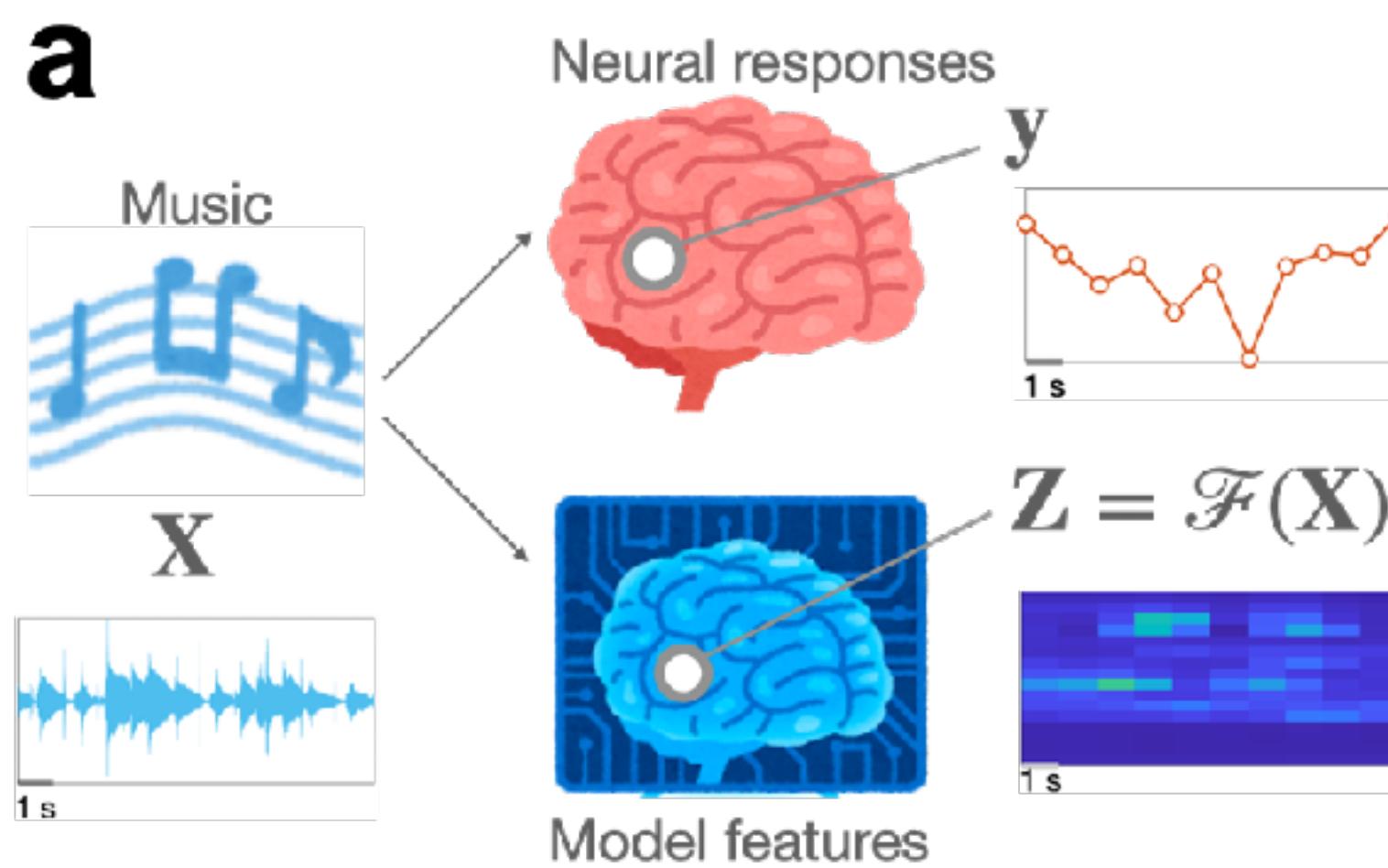
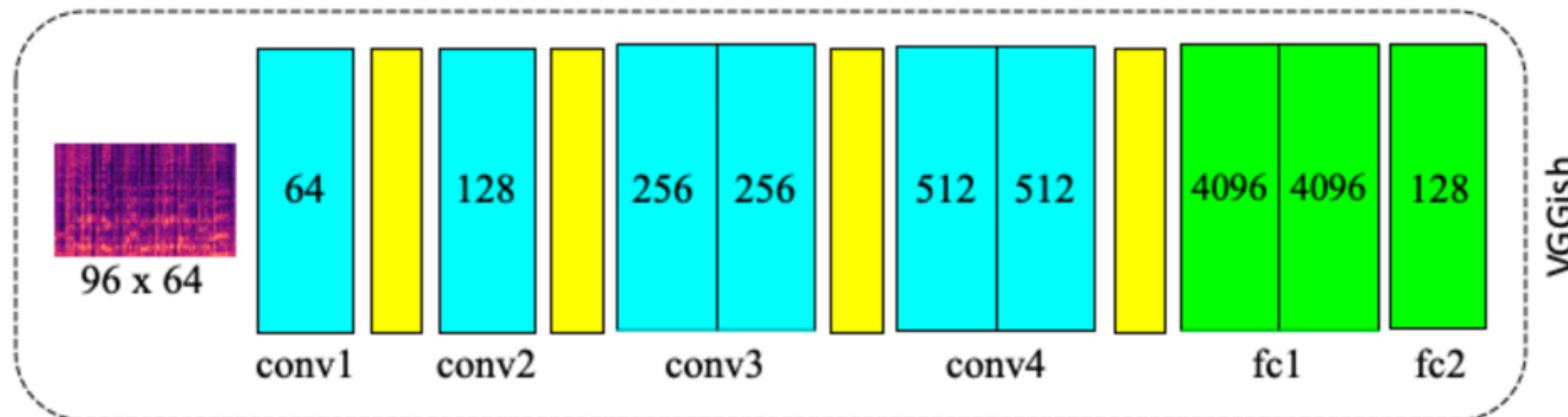
Temporal response function of melodic surprisal

Non-invasive recording of EEG (scalp potential) D



Representational gradient of musical emotions in fMRI

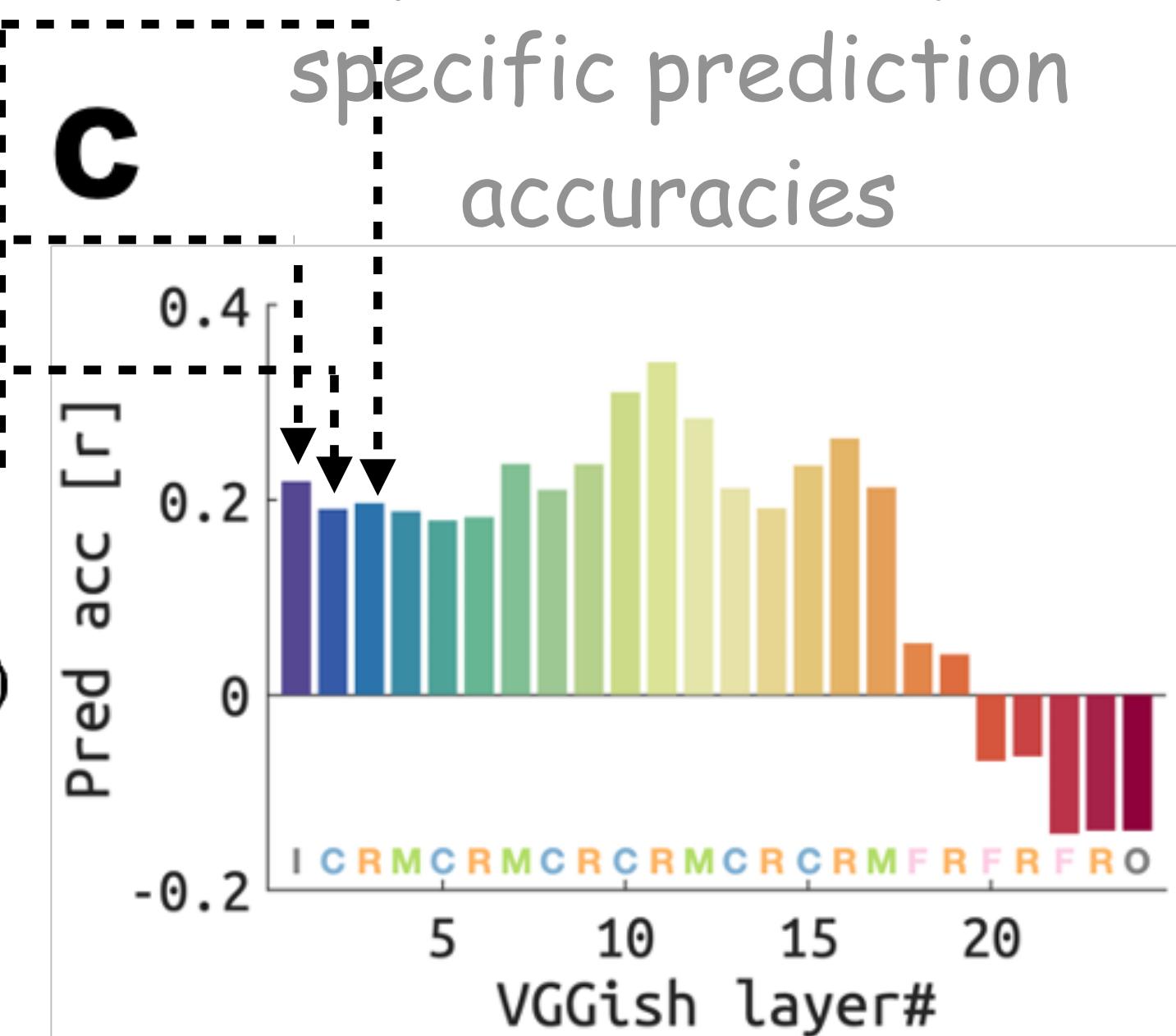
Non-invasive scanning of functional MRI (local paramagnetism)



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

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

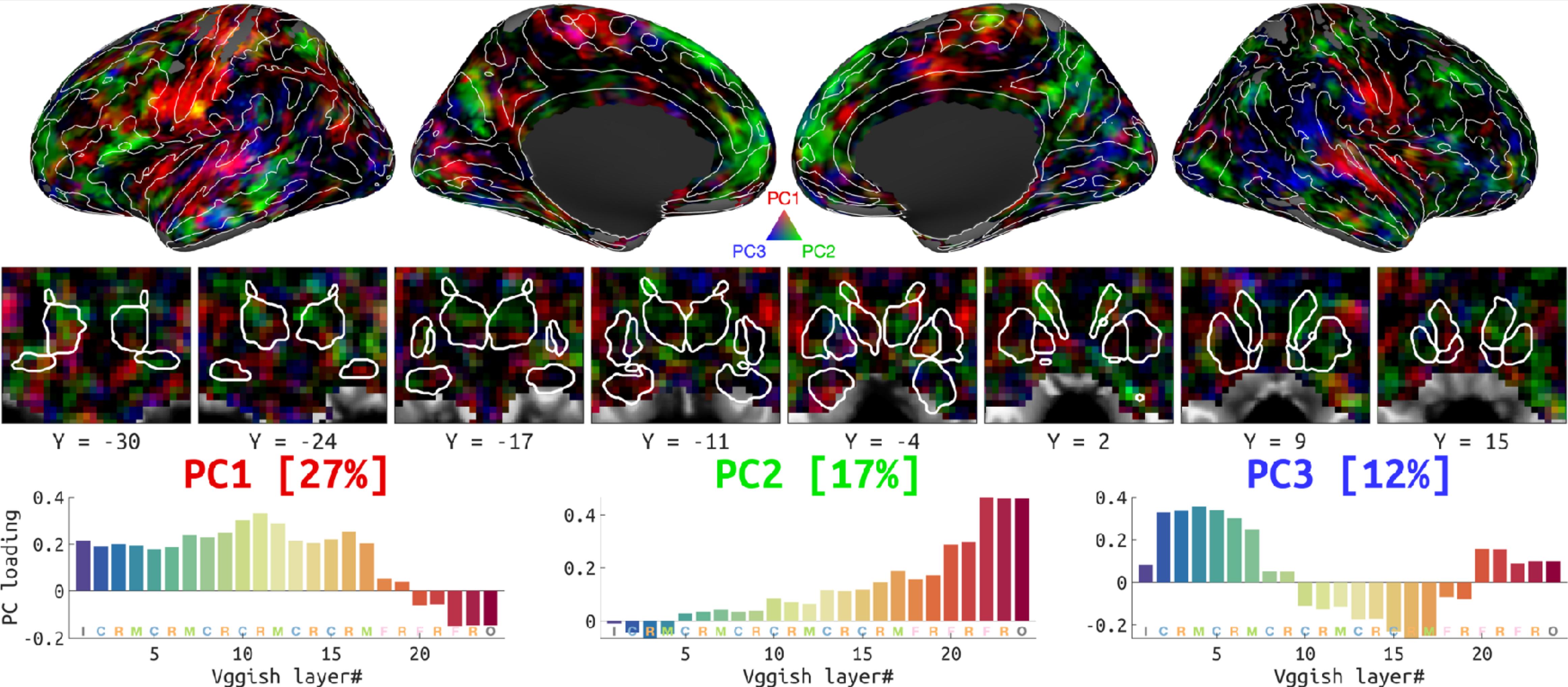
A "profile" of layer-specific prediction accuracies



Kim et al., In Prep.

Representational gradient of musical emotions in fMRI

Non-invasive scanning of functional MRI (local paramagnetism)



Summary: Analysis

How to analyze naturalistic experiment

- Data-driven methods:
 - **Intersubject synchrony**
 - **Event structure analysis**
- Hypothesis-based methods:
 - **Linearized encoding analysis**

Any questions?

Analytic methods

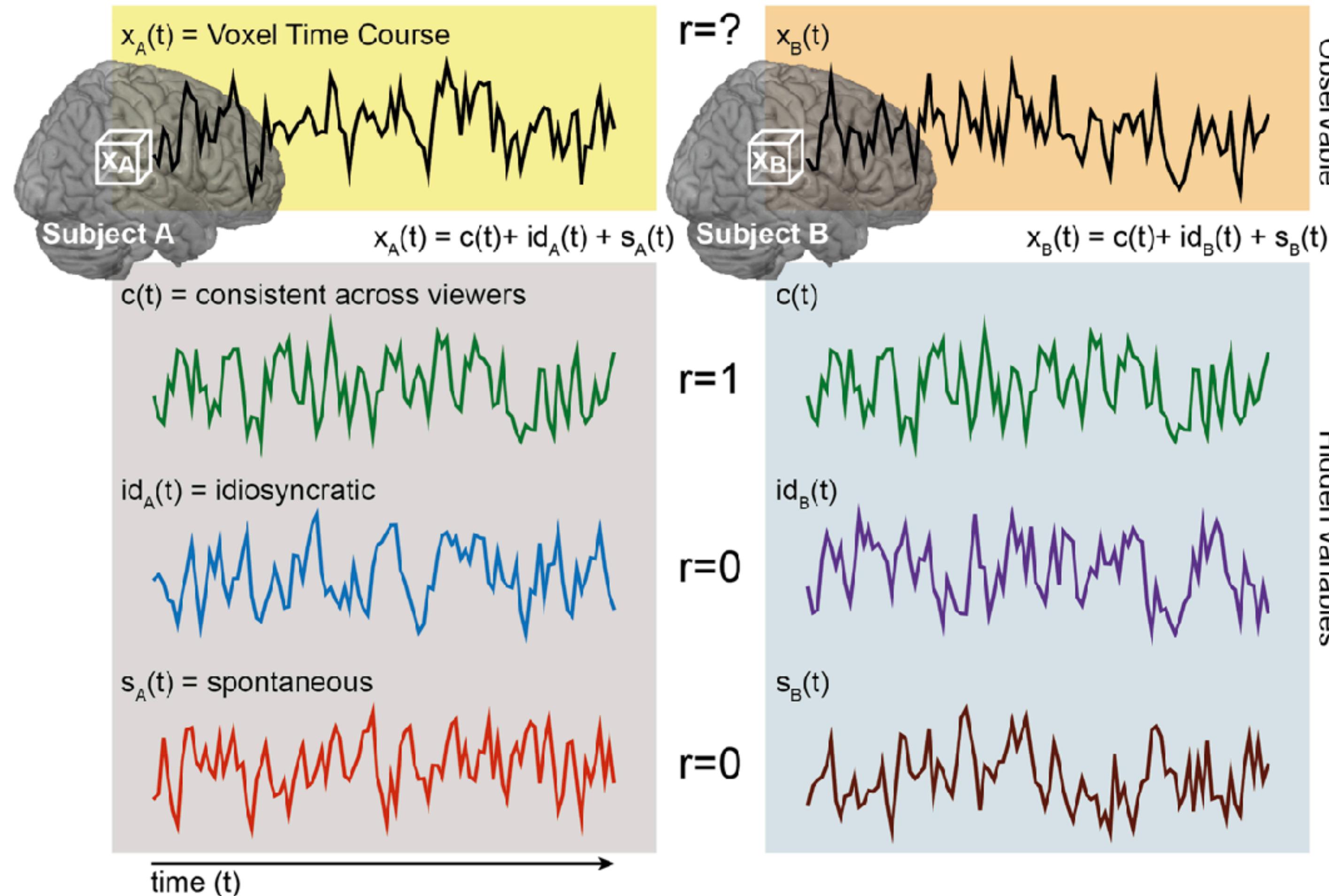


How to analyze naturalistic experiments

Hands-on

Let's try something simple first: ISC

$$x_A(t) = \alpha_A c(t) + \beta_A i d_A(t) + \varepsilon_A(t)$$



$$r_{AB} = r(x_A, x_B) \quad r_{AB}^2 \sim \alpha_A \bullet \alpha_B$$

$$\lim_{N \rightarrow \infty} (\bar{x}(t)) = c(t)$$

$$\because \mathbb{E}[id(t)] = 0, \mathbb{E}[\epsilon(t)] = 0$$

$$\hat{c}(t) = \bar{x}_{-i}$$

- **Intersubject correlation (ISC):** strength of stimulus-driven activity that is common across subjects
- No modeling of stimuli is required to compute ISC; but unspecific to actual activity (could be a **synchronized increase or decrease**)

Proposed pipeline

Time-resolved intersubject correlation

- **Stationarization of non-stationary time series**
- **Intersubject correlation calculations with a sliding window**
- **Statistical inference with sign-flipping**

Proposed pipeline

Time-resolved intersubject correlation

- **Stationarization of non-stationary time series**
- Intersubject correlation calculations with a sliding window
- Statistical inference with sign-flipping

Stationarization of non-stationary time series

Why? Because correlation assumes stationary noise

- Polynomial detrending
- High-pass filtering
- Derivatives (if downsampled enough)
- (Or not, just compute frequency-wise coherence and ignore low-frequency parts)

Proposed pipeline

Time-resolved intersubject correlation

- **Stationarization of non-stationary time series**
- **Intersubject correlation calculations with a sliding window**
- **Statistical inference with sign-flipping**

ISC calculation

Pseudocode

- let $X = \text{sum}(x(j))$ for all subjects
- for i-th subject:
 - $r(i) = \text{corr}(x(i), X - x(i))$

Time-resolved ISC calculation

Pseudocode

- **for** j-th time window:
 - **let** $X = \text{sum}(x(i, \text{TW}(j)) \text{ for all subjects}$
 - **for** i-th subject:
 - $r(i, j) = \text{corr}(x(i, \text{TW}(j)), X - x(i, \text{TW}(j)))$

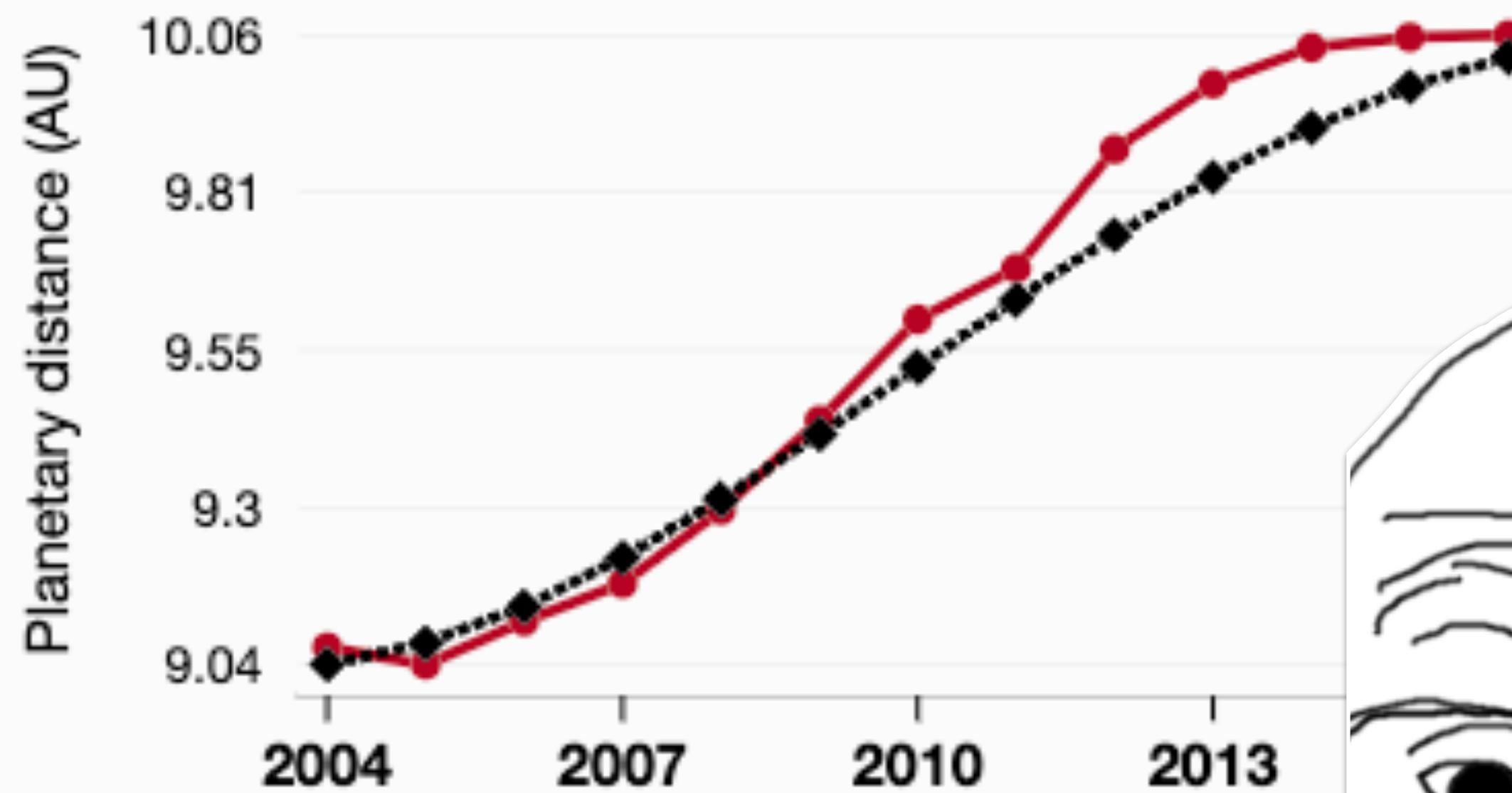
: Still, how do we know if this is not due to the random noise in the data?



The distance between Saturn and the Sun

correlates with

Google searches for 'how to'



◆ The average distance between Saturn and the Sun (black diamonds) · Source: Calculated using Astropy's solar system module · Data: NASA/JPL Solar System Dynamics

● Relative volume of Google searches for 'how to' (red circles) · Source: Google Trends · Data: Google Trends · Google Trends uses a scale from 0 to 100, where 100 is the maximum search volume for a query in a given month

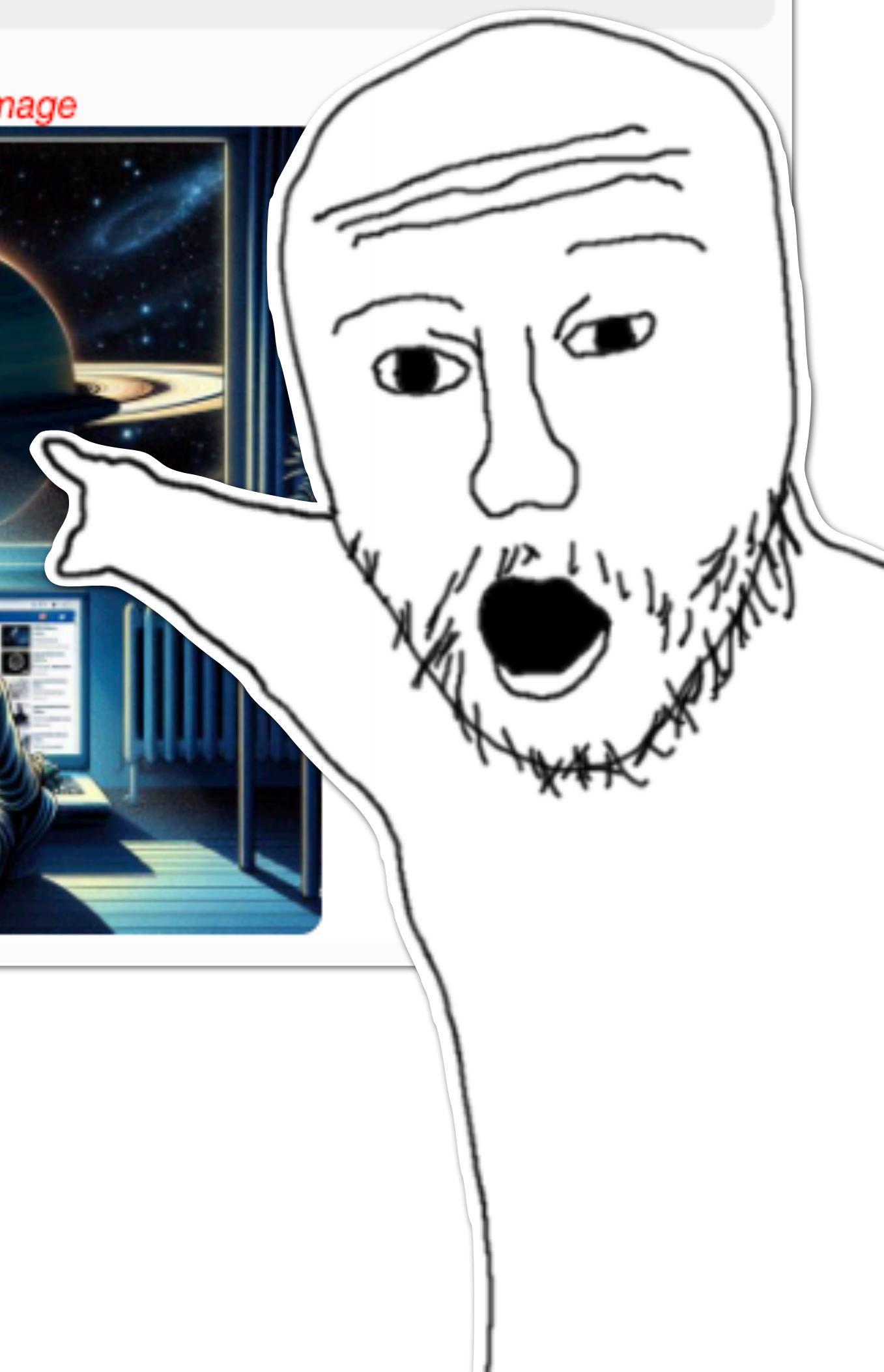
2004-2023, $r=0.967$, $r^2=0.935$, $p<0.01$ · tylervigen.com/1522

Ringed Planet's Orbit and Desire to Procreate: A Statistical Analysis

Show GenAI's made-up explanation

Fewer distractions from Saturn's bling led to some out-of-this-world romance. Looks like love isn't the only thing that's been cosmic lately!

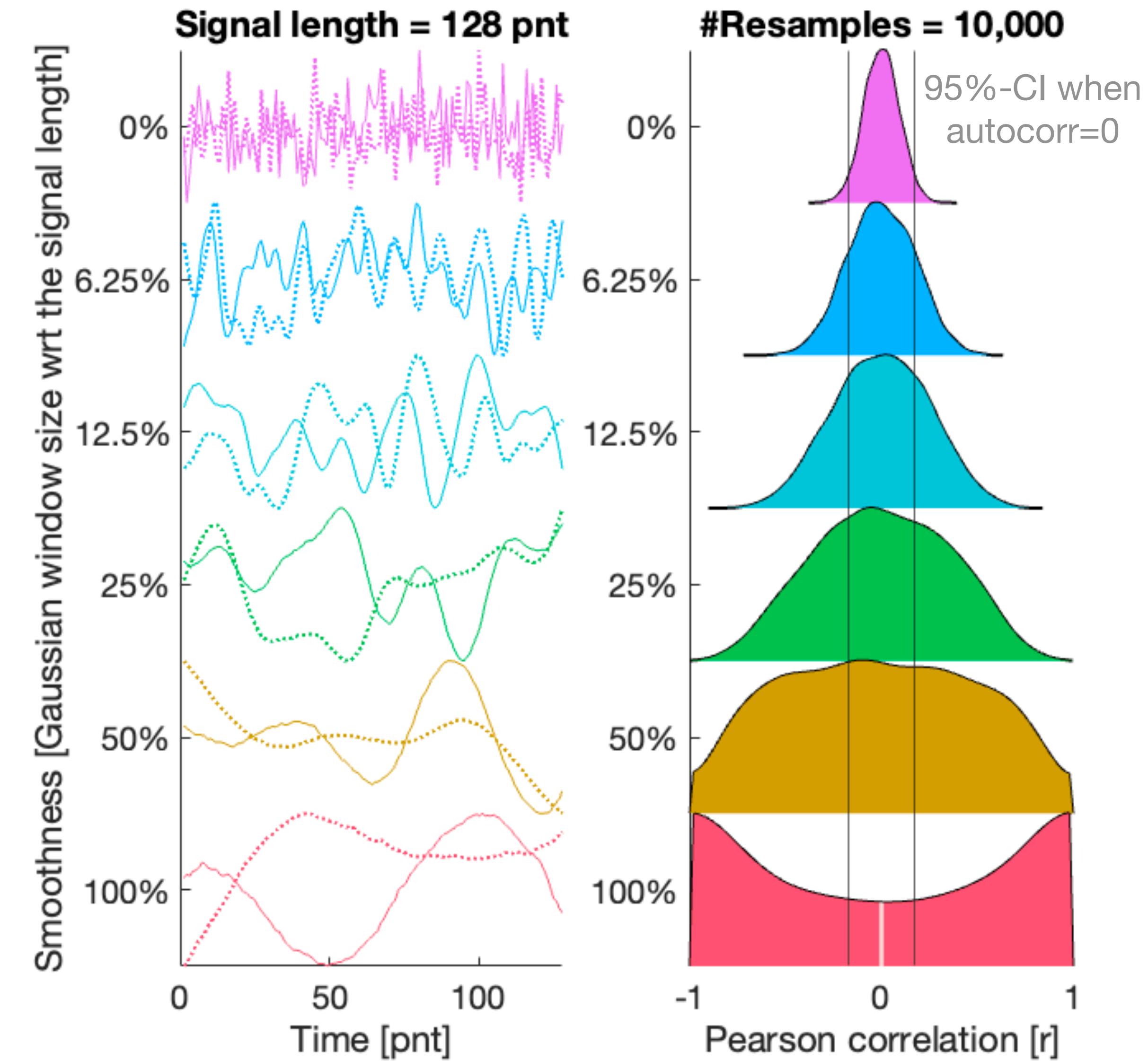
Show GenAI image



Statistical inference

Tests for time series

- Many **time series** look alike (i.e., spurious correlation) due to auto/serial-correlation.
- **Non-parametric statistical tests are feasible for any arbitrary noise distribution** with "proper" null (surrogate) data.
- Explicitly estimation autocorrelation from finite (non-infinite) data is hard.
- Preserving the dependency structure (e.g., serial/spatial, ...) is rather easy.
- For a signed time series, simply randomizing its polarity (sign-flipping) would preserve serial correlation structure.



Proposed pipeline

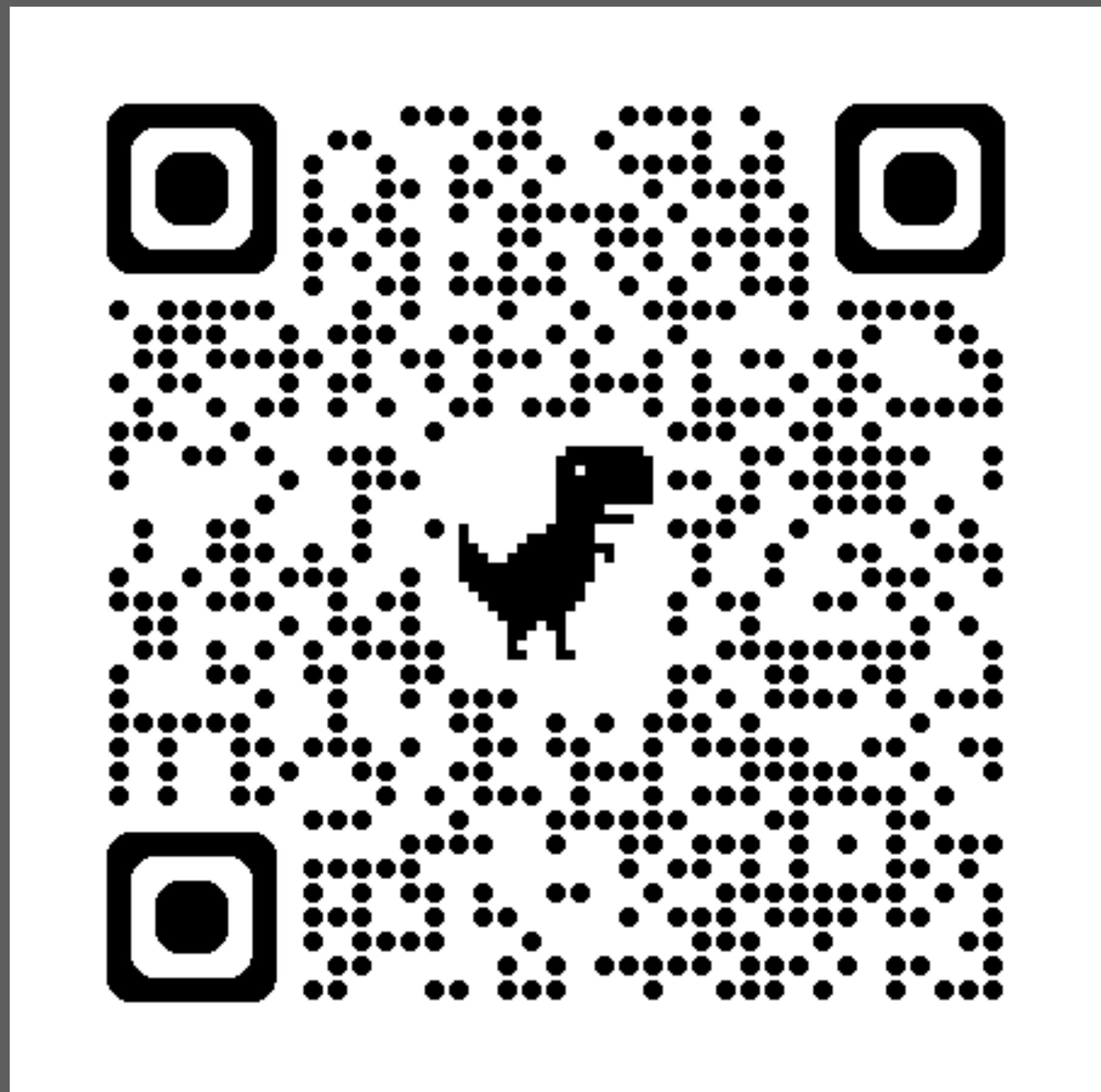
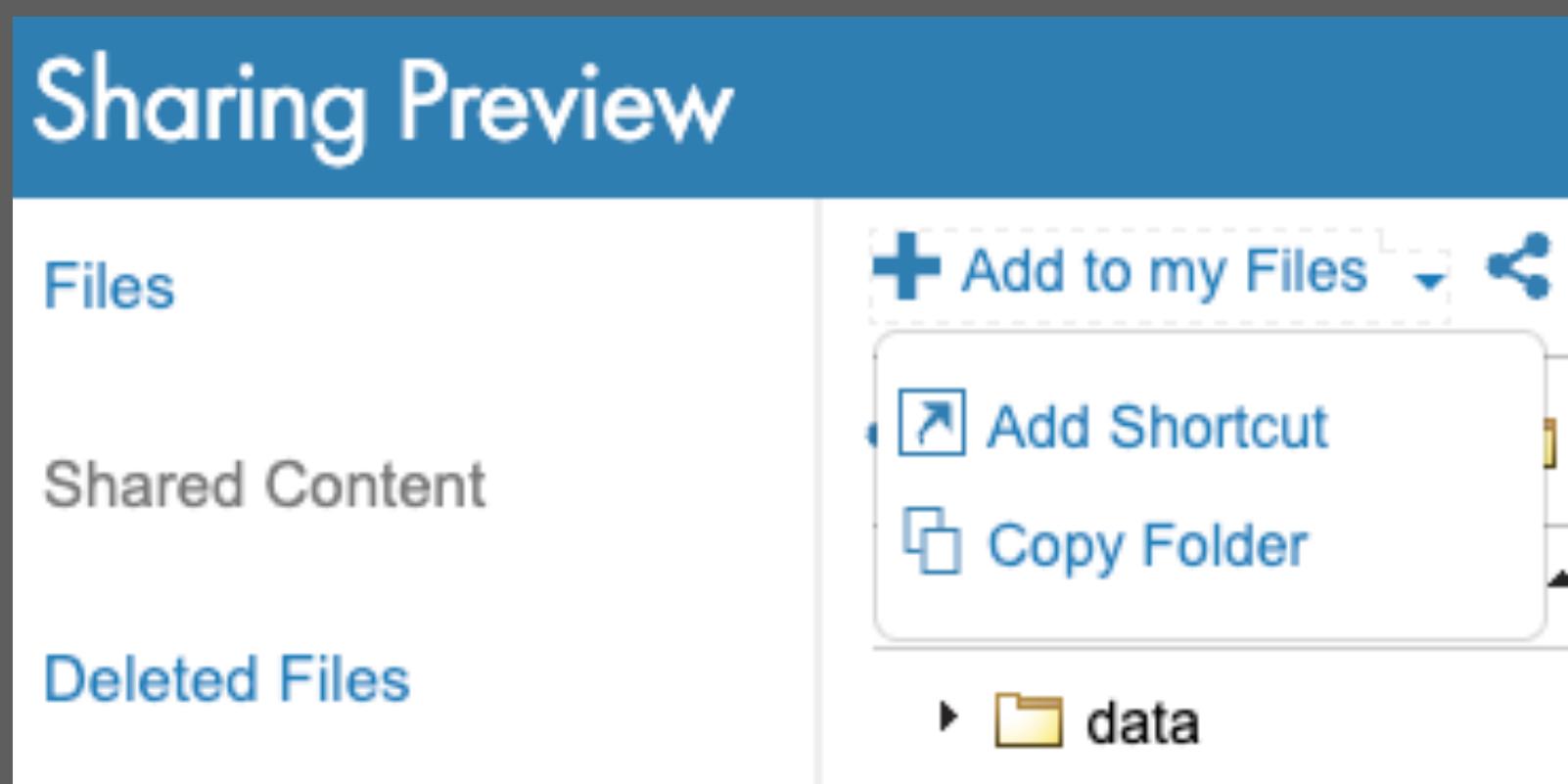
Time-resolved intersubject correlation

- **Stationarization of non-stationary time series**
- **Intersubject correlation calculations with a sliding window**
- **Statistical inference with sign-flipping**

MATLAB ONLINE

How to add the SHARED drive to your account

1. log in <https://MATLAB.mathworks.com>
2. Scan the T-rex or click on the share link [<https://drive.mathworks.com/sharing/1f26fcd8-c113-4efc-8f07-f7b480b4b74d>]
3. Clink on "Add to my Files" and "Copy Folder"



(Would we have time for this?)

Linearized Encoding Analysis (LEA)

- **Extracting music features**
- Preprocessing of non-stationary time series
- **Modeling neural delays using finite impulse response**
- **Ridge regression with cross-validation**
- Statistical inference with smooth time series
- https://seunggookim.github.io/teaching/teach_w_ksmpc24/

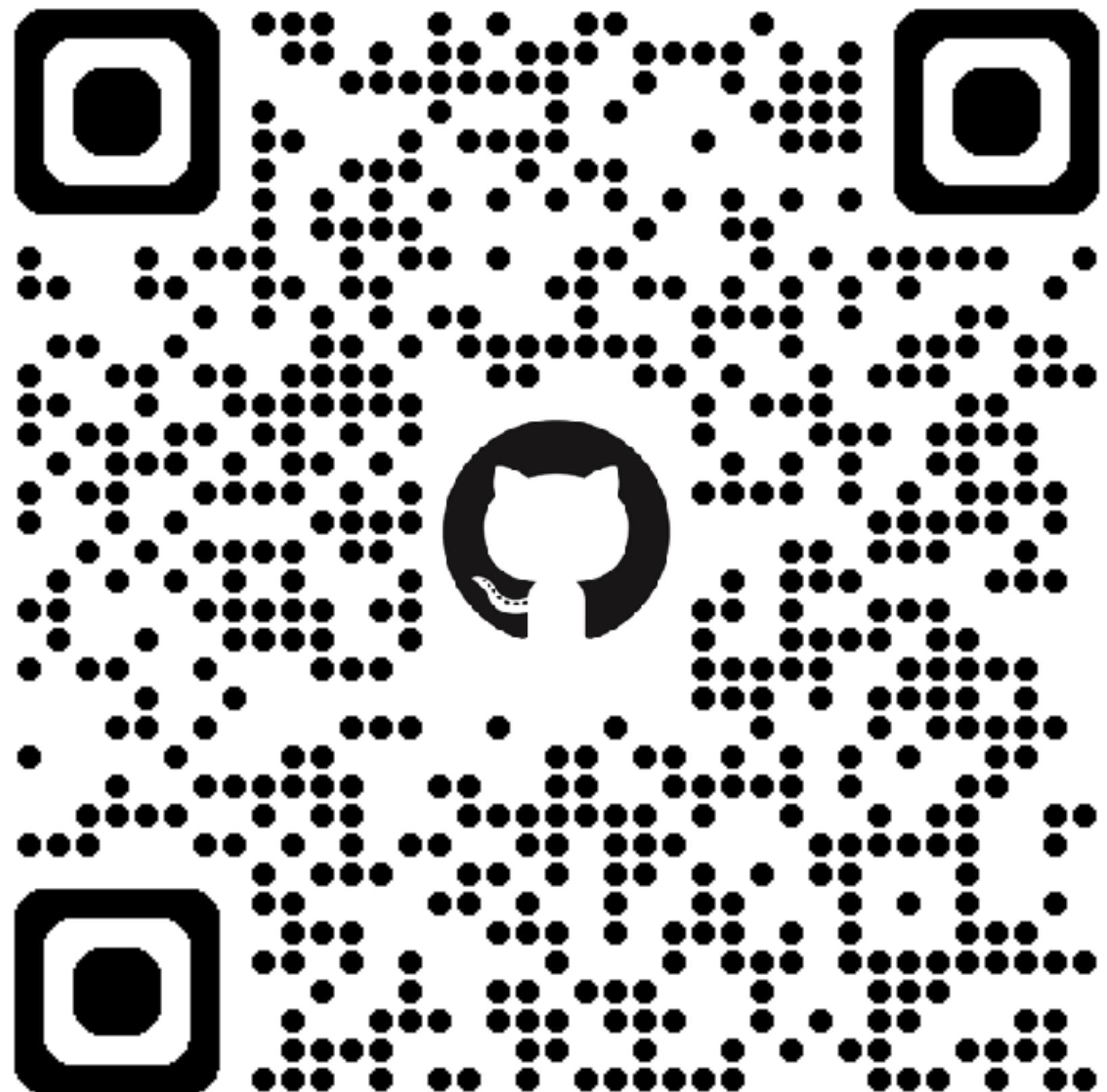
Summary of the session

Music, Brain, and Emotions

- Music may evoke intense emotions while the underlying mechanisms remain unclear.
- Behavioral/physiological responses to **naturalistic music** can be analyzed using novel methods and computational models.
- **Deep phenotyping** of musical emotions may help to better understand how musical information is transformed to evoke emotions.

Resources

<https://github.com/seunggookim/mpscog-ca25-mbe>



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



Thank you for your attention and participation! 😎