

How Music Alters Decision Making: Impact of Music Stimuli on Emotional Classification

†Elad Liebman, †Peter Stone, ‡Corey N. White

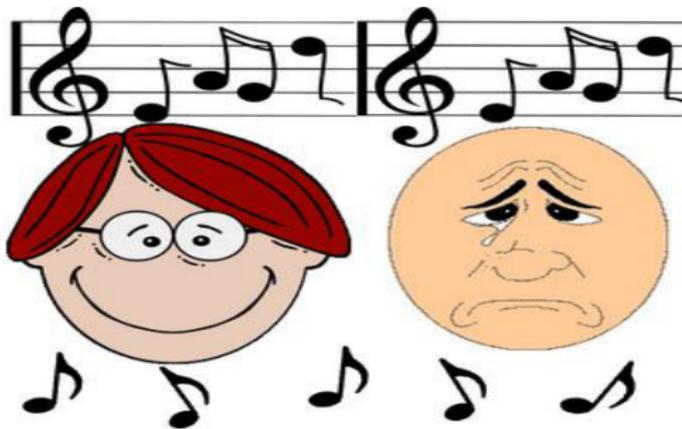
†Computer Science Department
The University of Texas at Austin

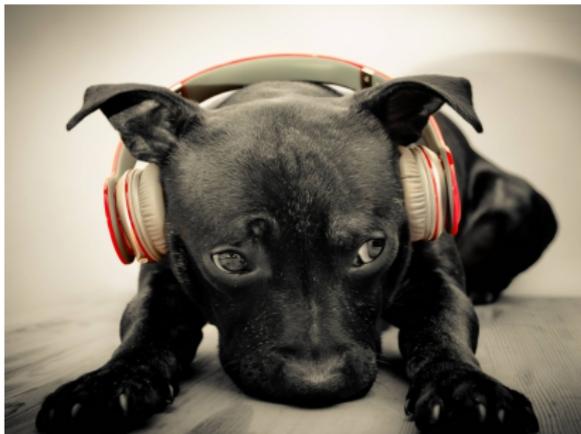
‡Department of Psychology
Syracuse University

October 29, 2015

Overview

- ▶ **Mood** can affect **emotional processing** (Behen 2011, Kuhbandner and Pekrun 2013).
- ▶ We explored how decision making is affected by **music**.
- ▶ Experiment - classify words as positive/negative while listening to music.
- ▶ Music chosen to induce mood.
- ▶ Results show music manipulation was **effective**.





- ▶ Robust evidence of **mood-congruent processing**, or bias (White et al. 2009, 2010).
- ▶ Music affects mood, but how does it that effect emotional decision making?

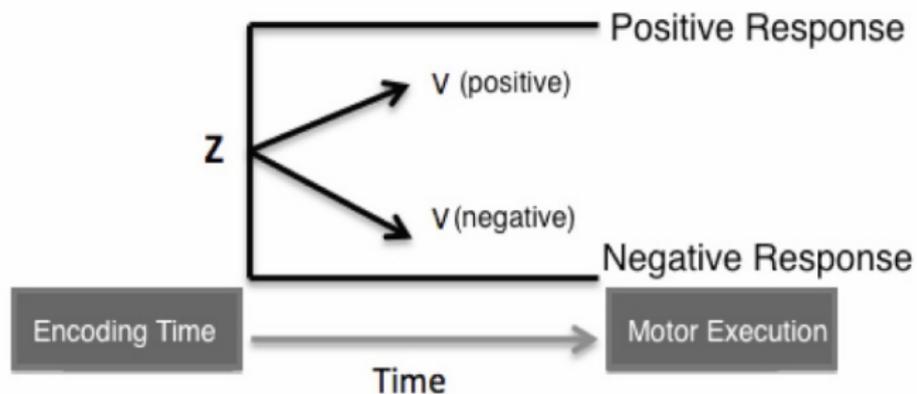
- ▶ People had to perform an **emotional task**.
- ▶ ...while listening to **music**.
- ▶ To analyze the results, we use a **Drift Diffusion Model (DDM)**(Ratcliff & McKoon, 2008).
- ▶ The DDM differentiates two types of bias:
 - ▶ Due to an **a priori preference**
 - ▶ Due to a **shift** in how stimuli are evaluated
- ▶ Model has been used in the past, but not in this context.

- ▶ Participants were shown **words** and asked to **classify** them.
- ▶ Words taken from a previous paper (White et al. 2013).
- ▶ Words were categorized into three categories:
 - ▶ Positive - success, happy
 - ▶ Neutral - shelves, sipped
 - ▶ Negative - worried, sad
- ▶ The task consisted of **4 blocks of 60 trials with 20 stimuli from each word condition.**

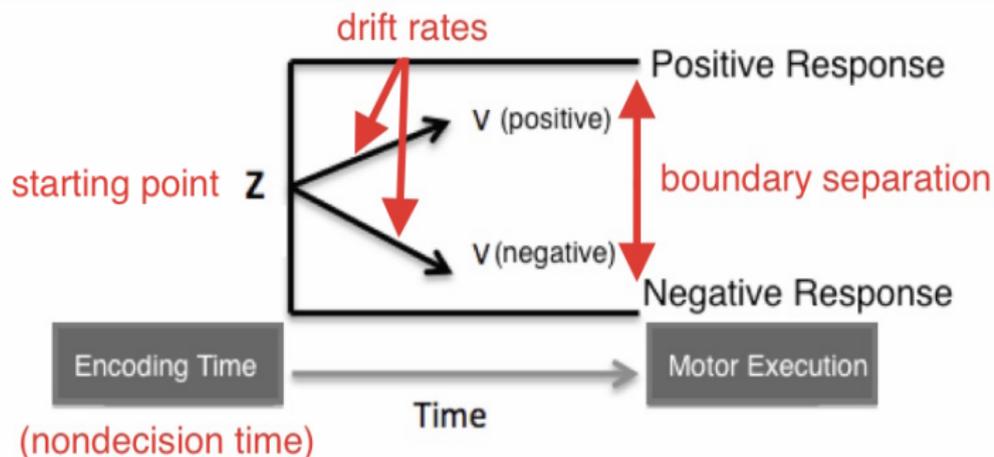
- ▶ A different song was played during each block.
- ▶ Words were randomly assigned (20 of each type).
- ▶ The DDM was fitted to each participant's data via convex optimization (minimizing χ^2).

Experiment console example:

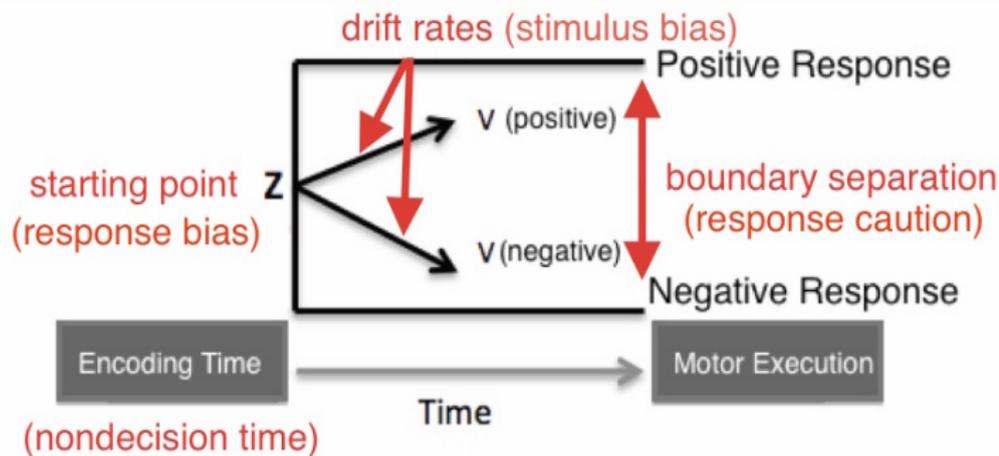
Drift Diffusion Model



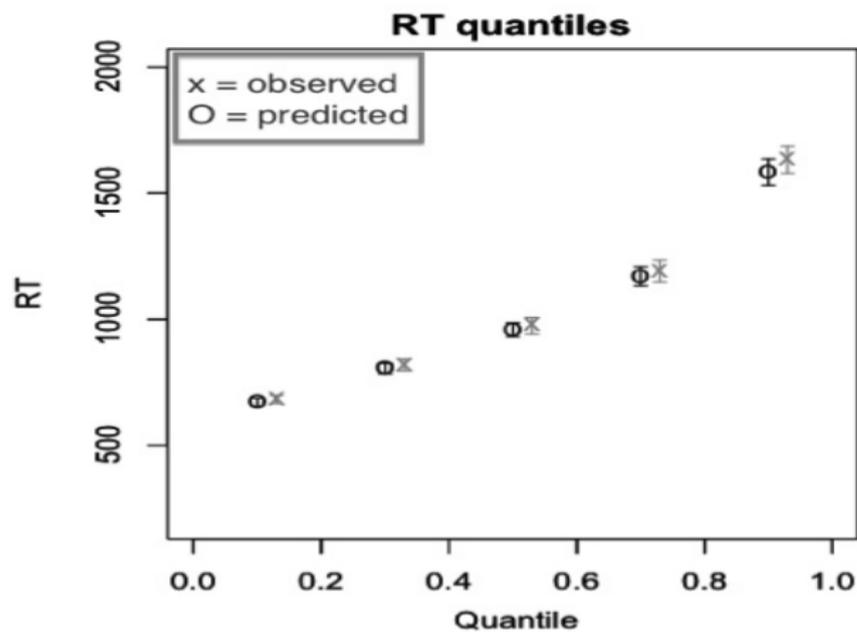
Drift Diffusion Model



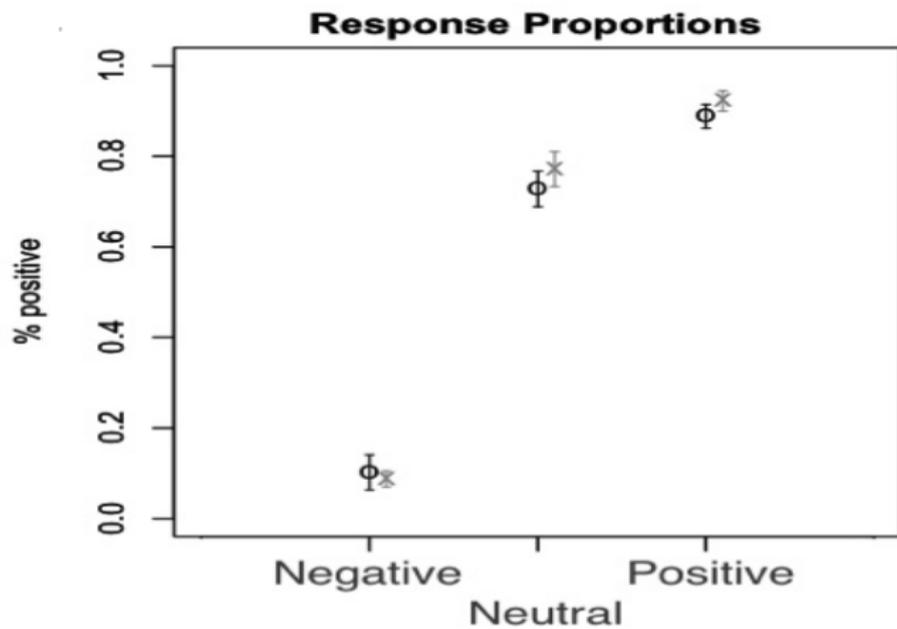
Drift Diffusion Model



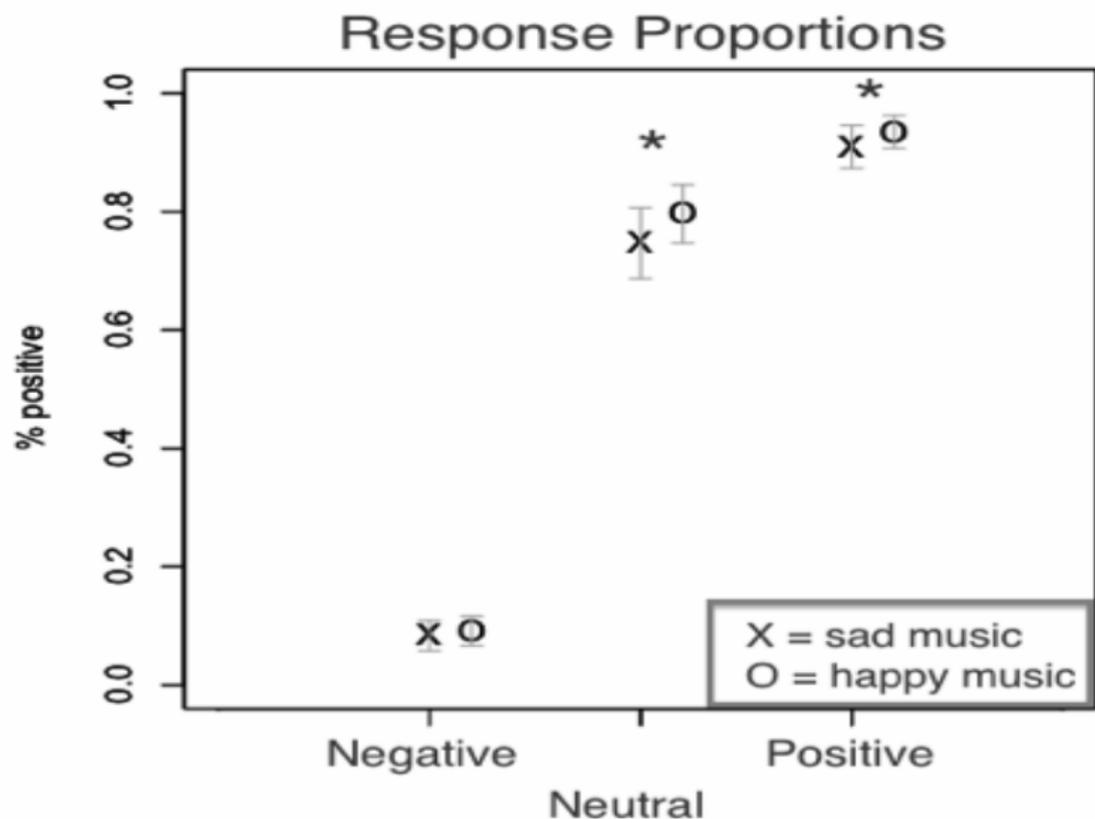
Model Fitting Quality



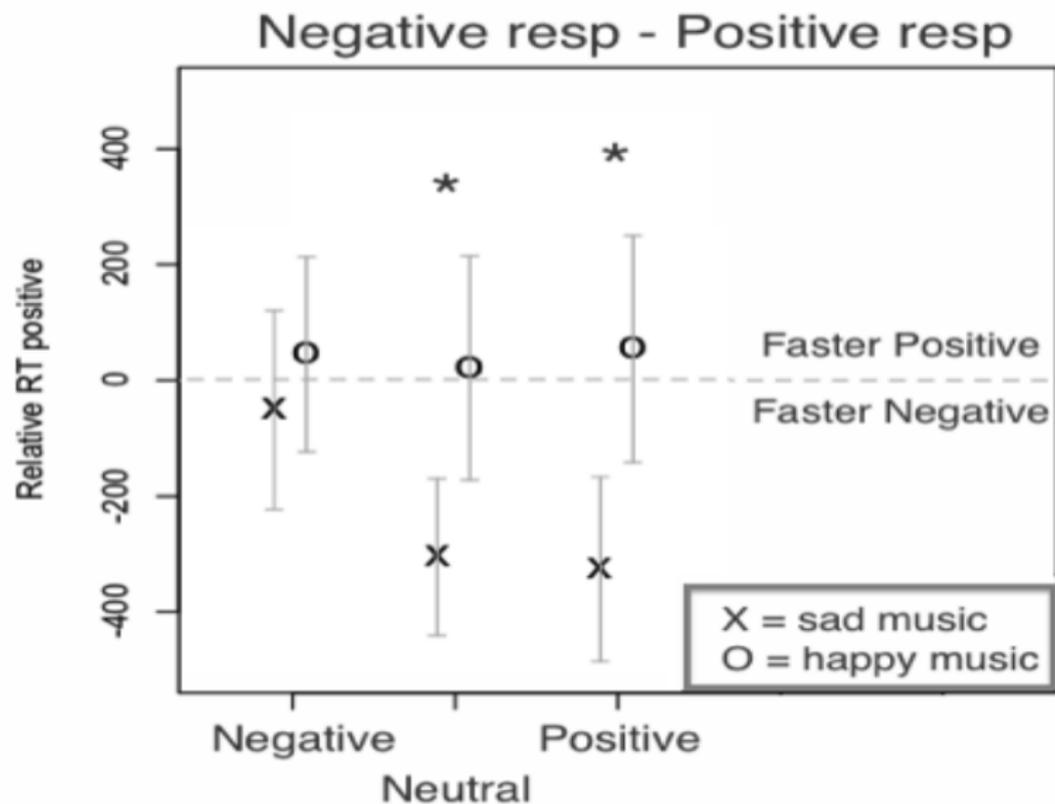
Model Fitting Quality



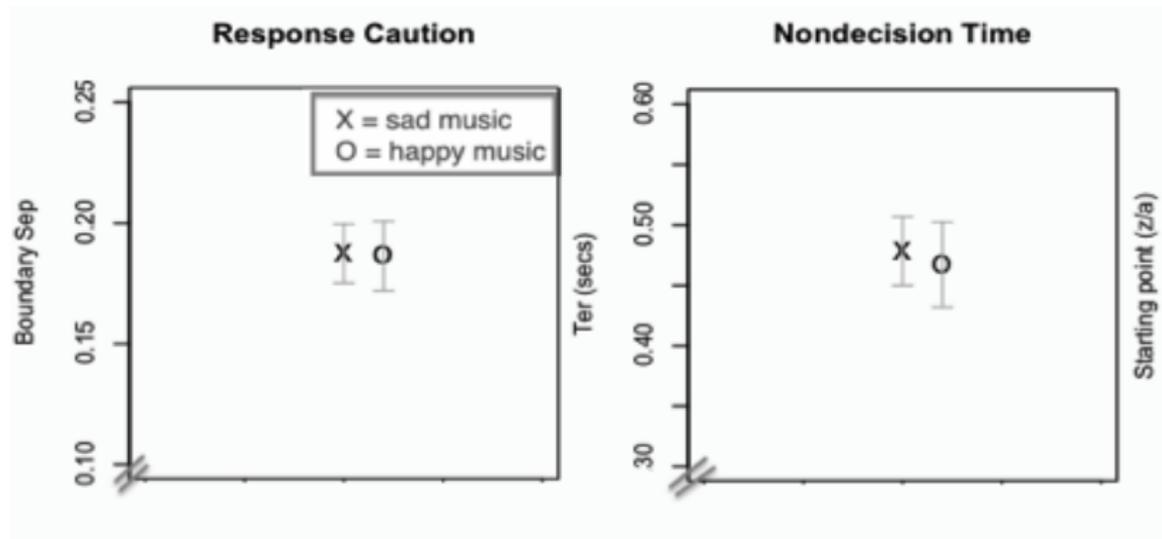
- ▶ Mood-induction successfully affected emotional bias.
- ▶ Happy music led to more “positive” responses overall.
- ▶ For the starting point, there was a significant shift.
- ▶ No reliable effect on the drift rates, response caution or nondecision time.
- ▶ Music affected the bias, but not the stimulus evaluation.



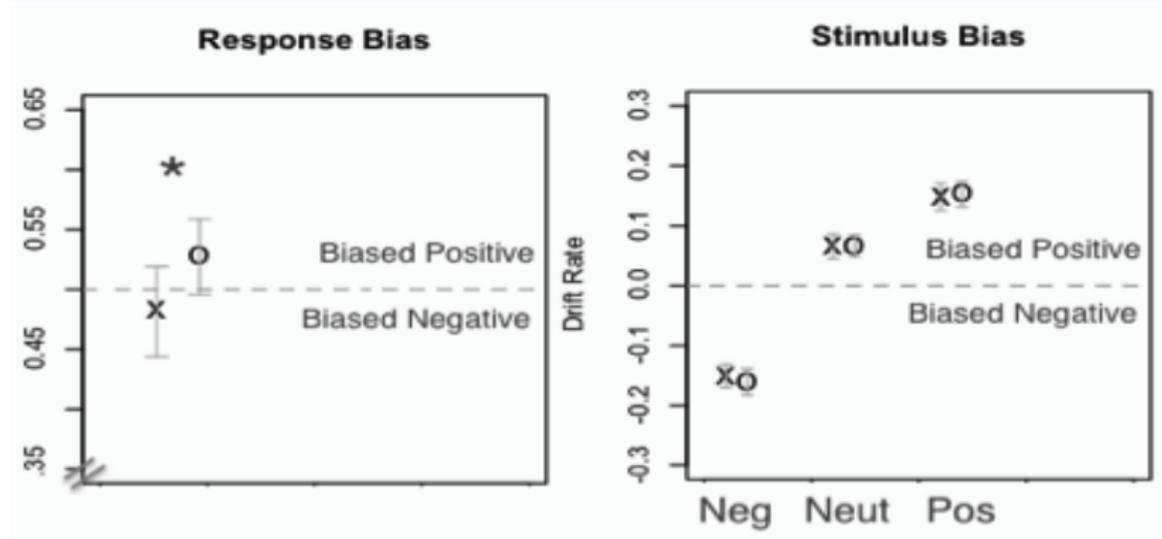
Results



Results



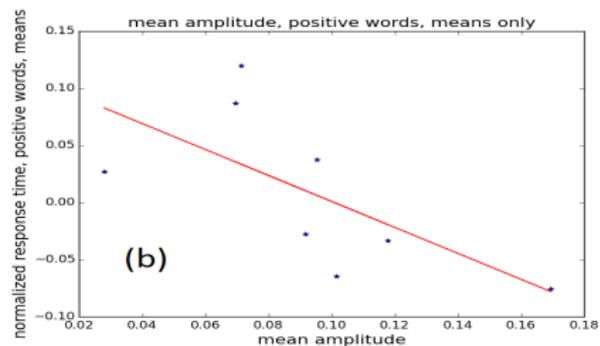
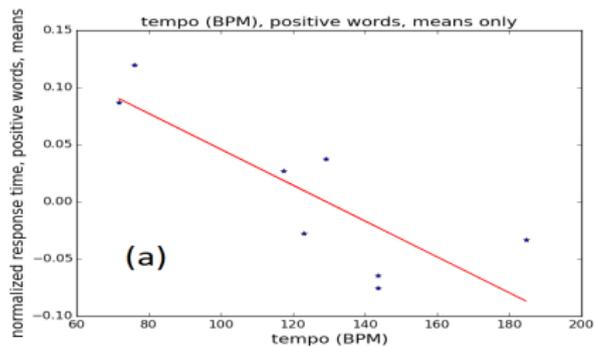
Results



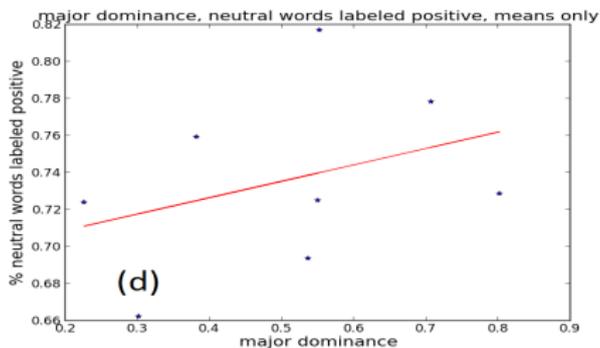
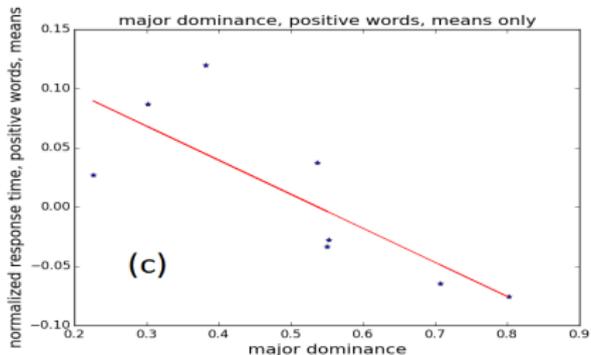
Analyzing Individual Auditory Features

- ▶ The partition to “positive” and “negative” is arbitrary.
- ▶ How do **specific music aspects** affect response patterns?
- ▶ We considered the 8 musical segments used.
- ▶ We focused on **tempo, loudness and major/minor ratio**.
- ▶ Studied how these features correlated with responses.

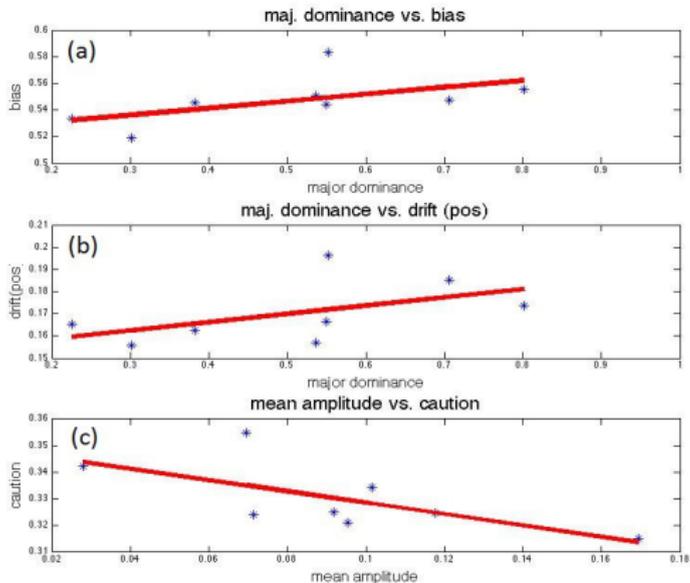
Analyzing Individual Auditory Features



Analyzing Individual Auditory Features



Analyzing Individual Auditory Features



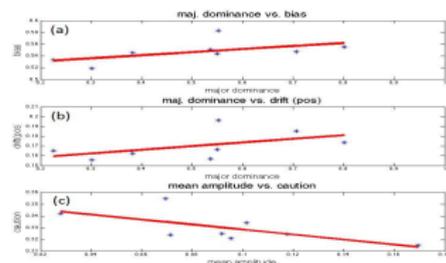
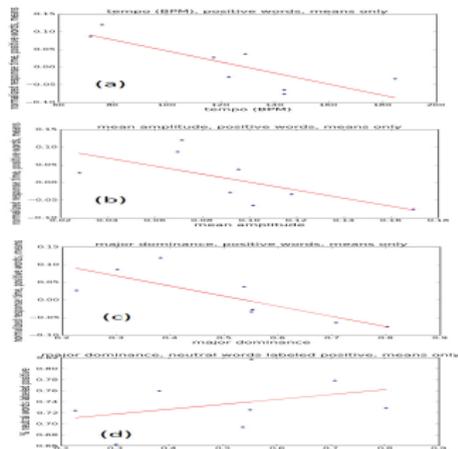
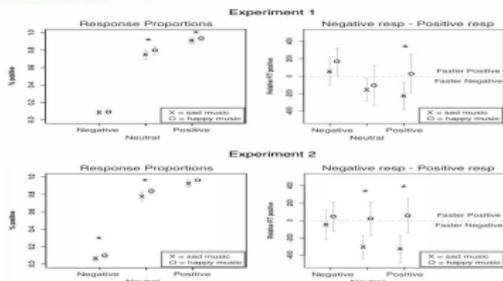
- ▶ Music does not significantly affect content evaluation.
- ▶ Rather, music **initializes a prior preference**.
- ▶ In other words, negative words stay as negative.
- ▶ But music biases the selection process.
- ▶ This type of analysis is **only possible thanks to the DDM**.

- ▶ Does this bias originate in the frequent pairing of sad/happy content with sad/happy music, or is it an innate property? Hard to tell.

- ▶ (...but we can show music correlates with responses on a basic feature level)

- ▶ Want to understand how music affects processing.
- ▶ The **Drift-Diffusion model** decomposed behavior into meaningful constructs.
- ▶ Participants classified words as **positive or negative**.
- ▶ They did so while listening to mood-inducing music.
- ▶ Music-induced mood **affected expectancy**.
- ▶ ...but not stimulus evaluation, caution or encoding time.
- ▶ **Tempo, loudness and harmony are correlated with participant behaviors.**

Questions?



Thank you!