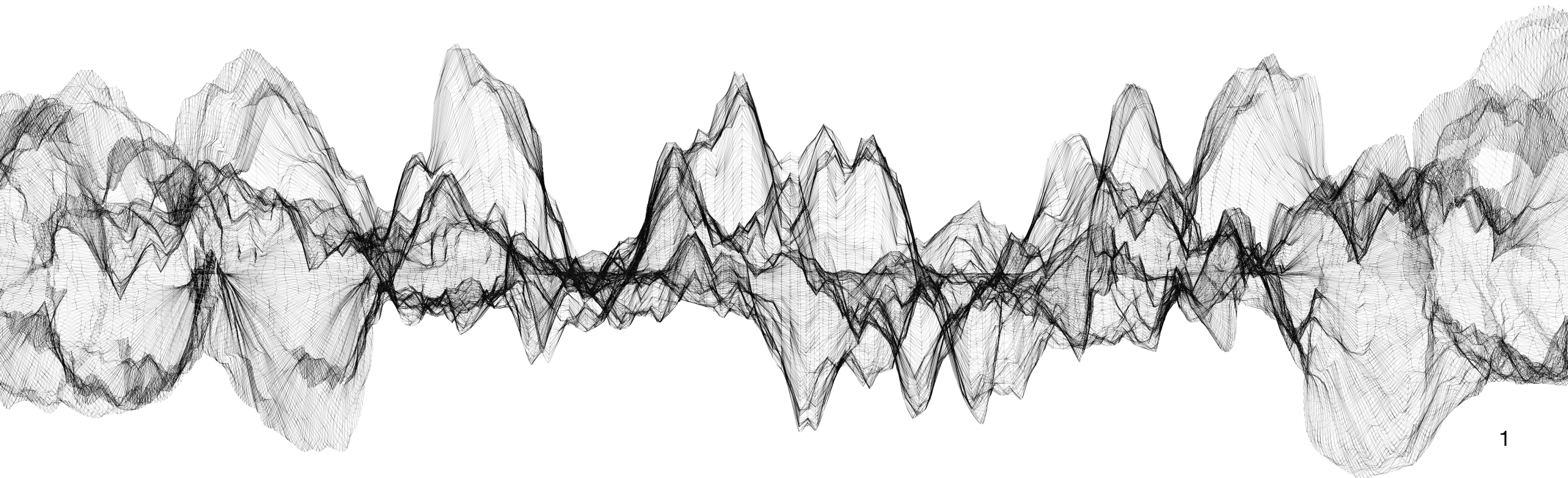




# Phonological (un)certainty weights lexical activation

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**Laura Gwilliams, David Poeppel, Alec Marantz & Tal Linzen**  
*7th January 2018*



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blind

big

bath

baptist

bond

band

ballot

book

b

break

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b a l

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balance

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# But what about ambiguity?

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- Real world speech is **noisy and ambiguous**; there is not a direct mapping between speech and phonemes

b b p p p

ballet

prove

pin

bath

pacify

bond

palate

book

beef

p

pants

balance

bind

paddle

boast

poke

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ballet

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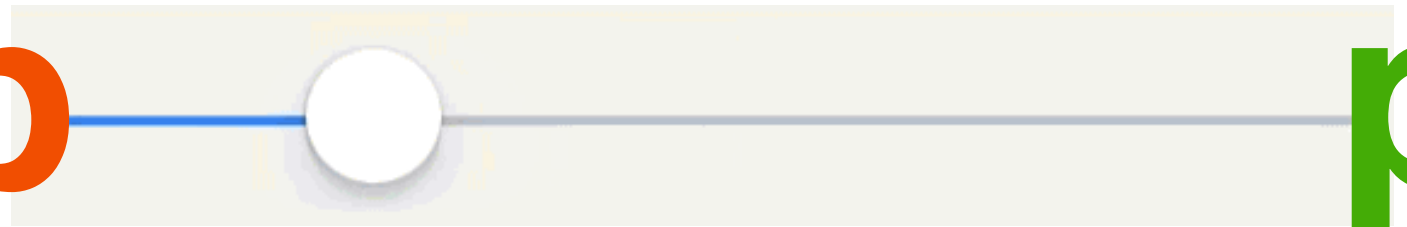
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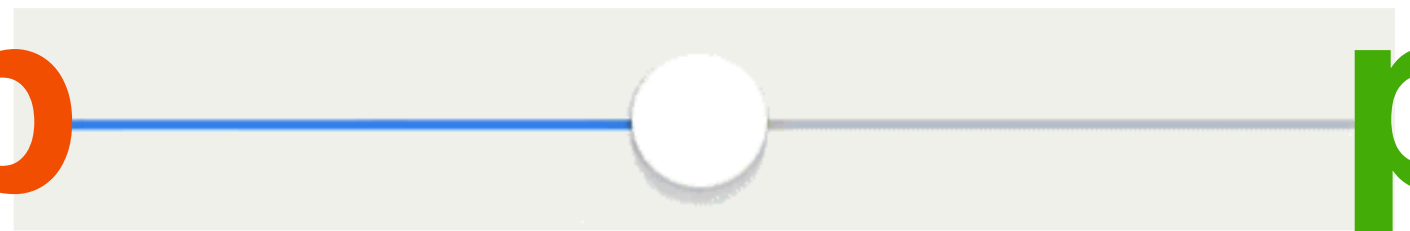
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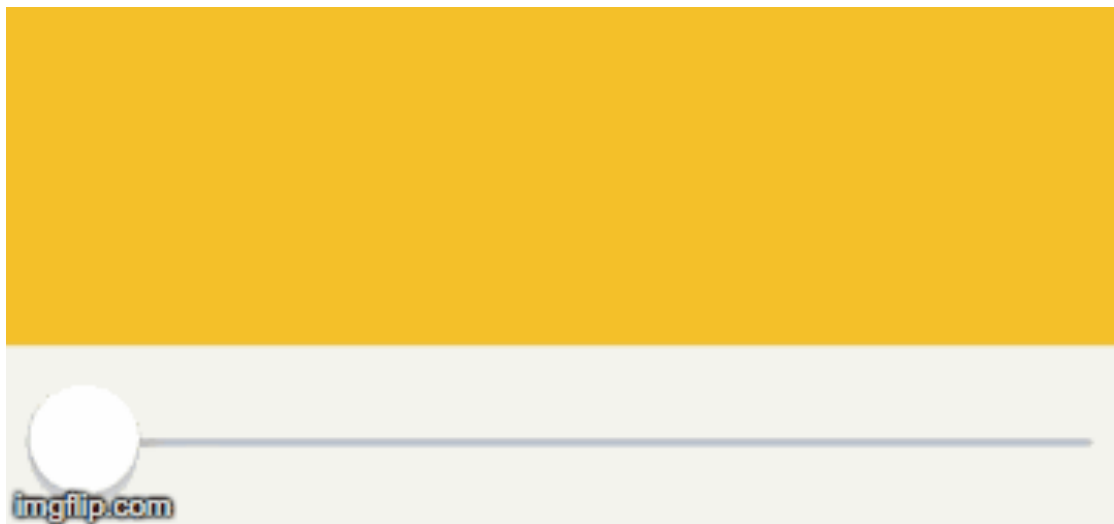
# Two Computational Models

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$$P(\varphi_a | A)$$

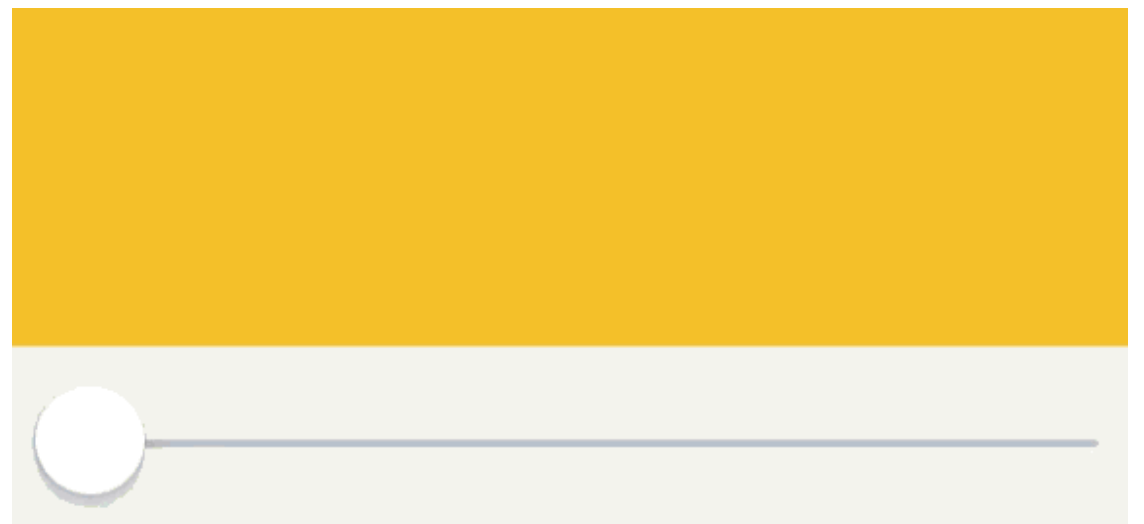
$\varphi_a$  = phoneme<sub>a</sub>       $A$  = acoustic input

## SWITCH-BASED



- 1 cohort of words
- binary acoustic term

## ACOUSTIC WEIGHTED



- 1+ cohort of words
- continuous acoustic term



# Research Question

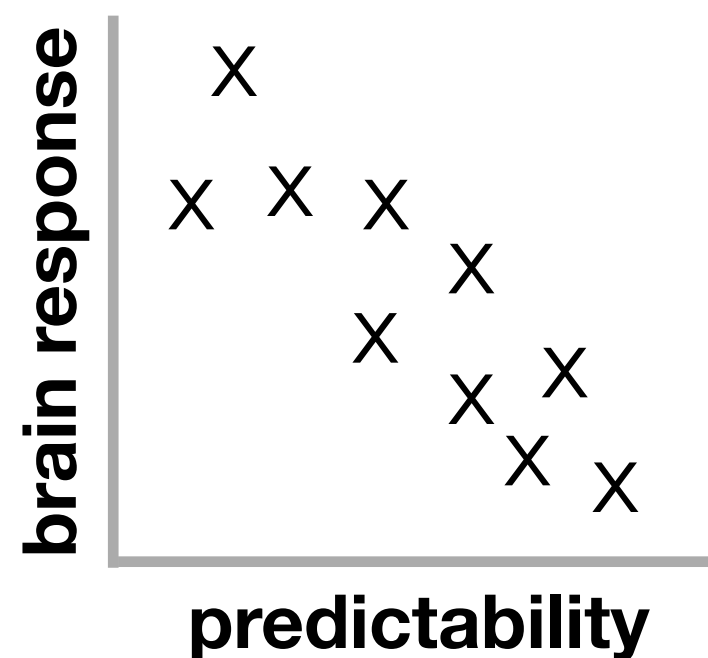
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Does acoustic-phonetic uncertainty  
weight activation at the lexical level?

# Prediction aids speech comprehension

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- The brain **predicts future linguistic content** in terms of phonemes, morphemes, words and syntactic structures
- When input is **predictable**, it is easier to process; reflected as a relative **reduction in neural amplitude**



# Quantifying predictability

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- **Surprisal:**

Probability of an outcome

$$-\log_2 \frac{f(\varphi_1, \dots, \varphi_t)}{f(\varphi_1, \dots, \varphi_{t-1})}$$

- **Entropy:**

Uncertainty over future input

$$-\sum_{w \in C} P(w|C) \log_2 P(w|C)$$

# Critical Variables

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- **Surprisal:**

Switch-based

Acoustic-weighted

$$-\log_2 \left( \underbrace{P(\varphi_a|A)}_{\text{red oval}} \frac{f(\varphi_a, \varphi_2, \dots, \varphi_t)}{f(\varphi_a, \varphi_2, \dots, \varphi_{t-1})} Q_a^t + \underbrace{P(\varphi_b|A)}_{\text{blue oval}} \frac{f(\varphi_b, \varphi_2, \dots, \varphi_t)}{f(\varphi_b, \varphi_2, \dots, \varphi_{t-1})} Q_b^t \right)$$

- **Entropy:**

Switch-based

Acoustic-weighted

$$P(w|C, A) = P(w|C_a) \underbrace{P(\varphi_a|A)}_{\text{red oval}} + P(w|C_b) \underbrace{P(\varphi_b|A)}_{\text{blue oval}}$$



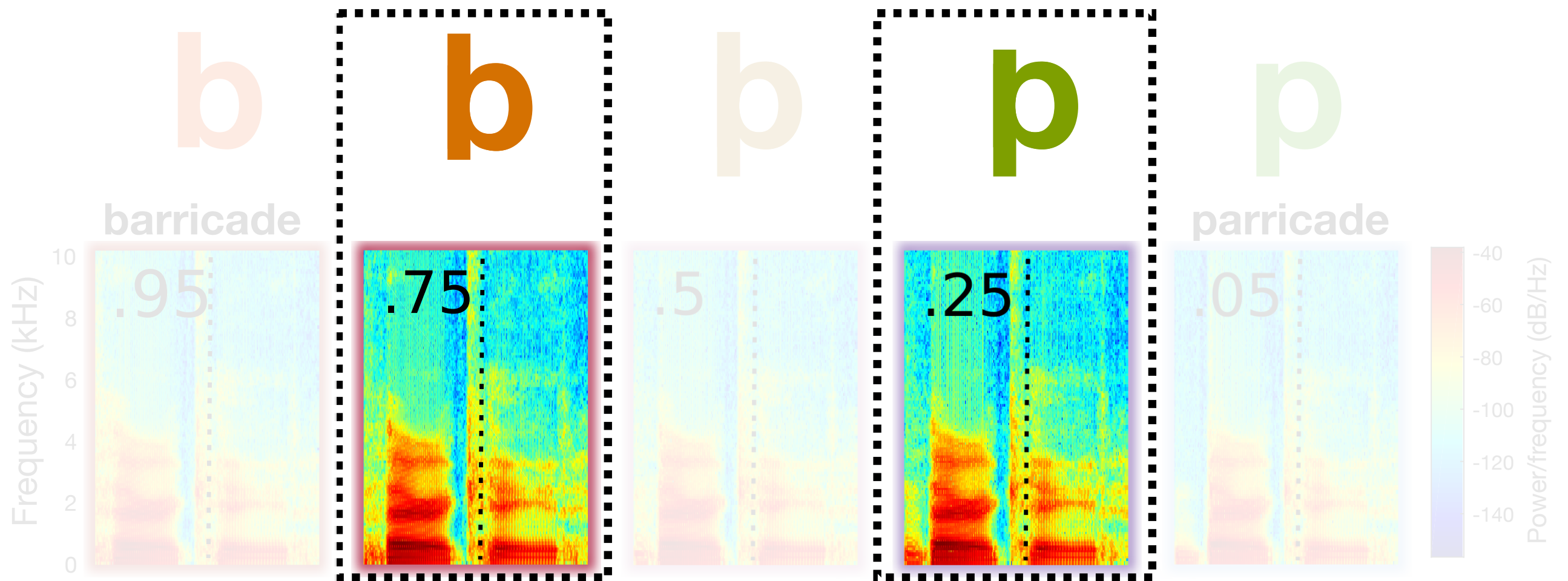
# Stimuli

Acoustic weighted:  $P(\varphi_a|A) = .75$

$P(\varphi_a|A) = .25$

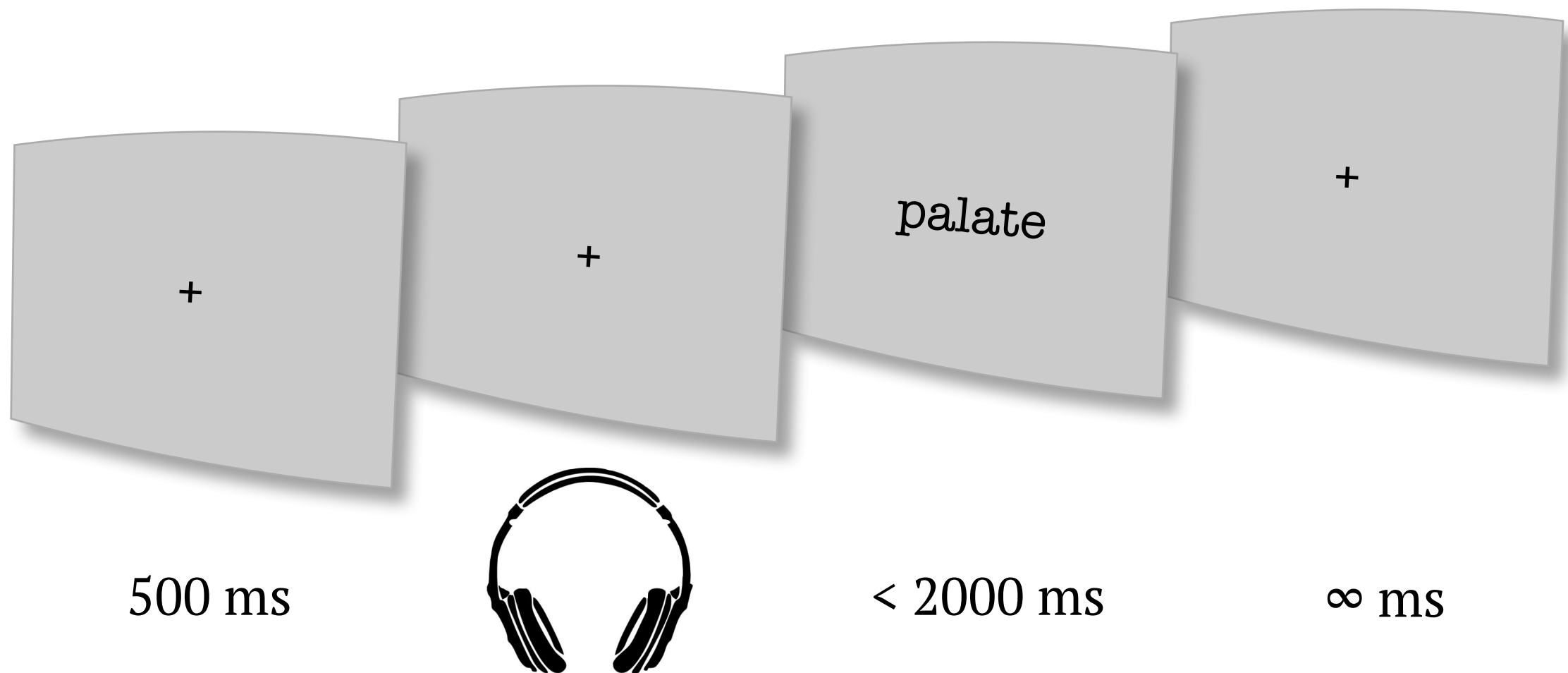
Switch-based:  $P(\varphi_a|A) = 1$

$P(\varphi_a|A) = 0$

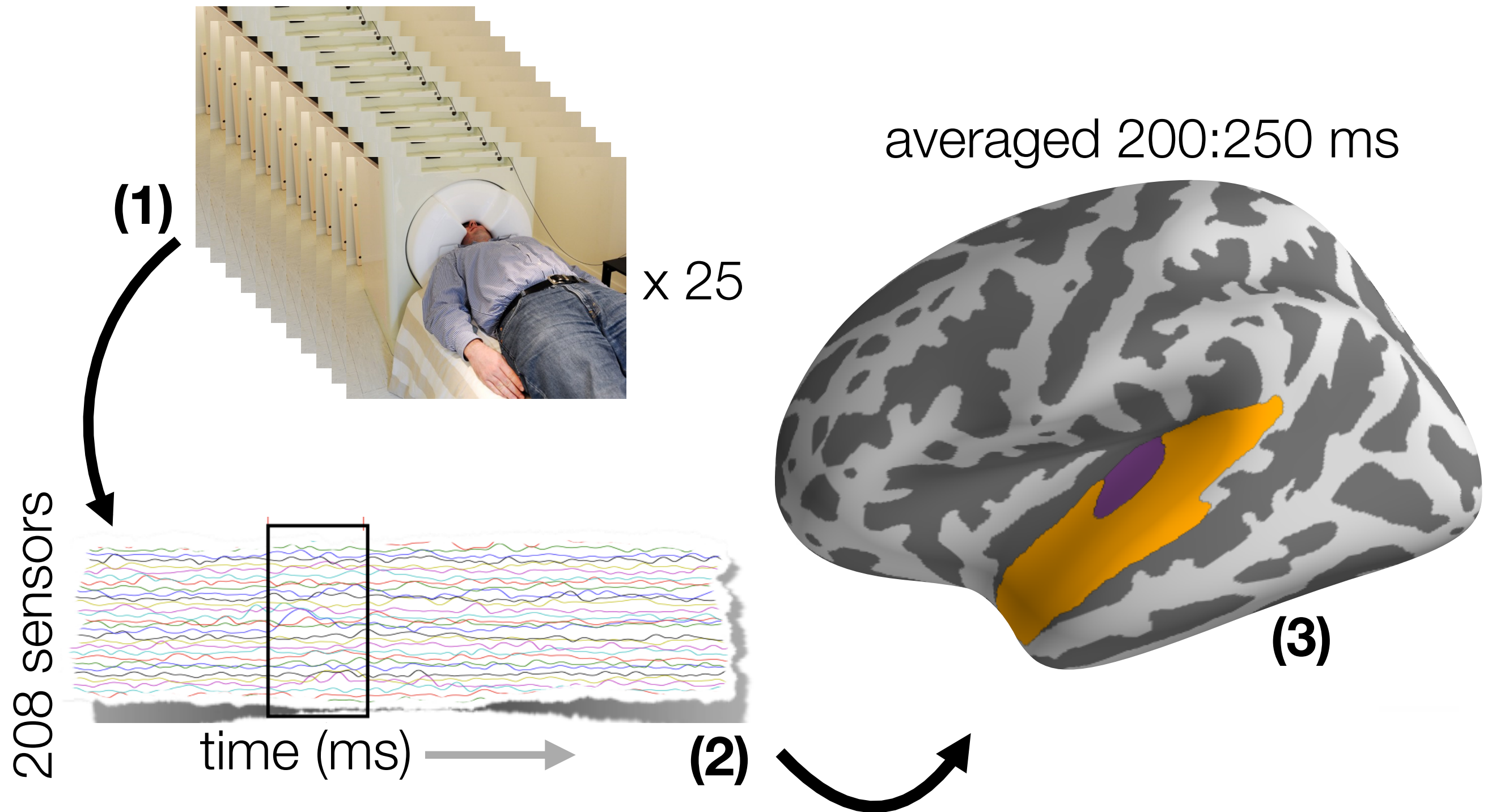


# Protocol

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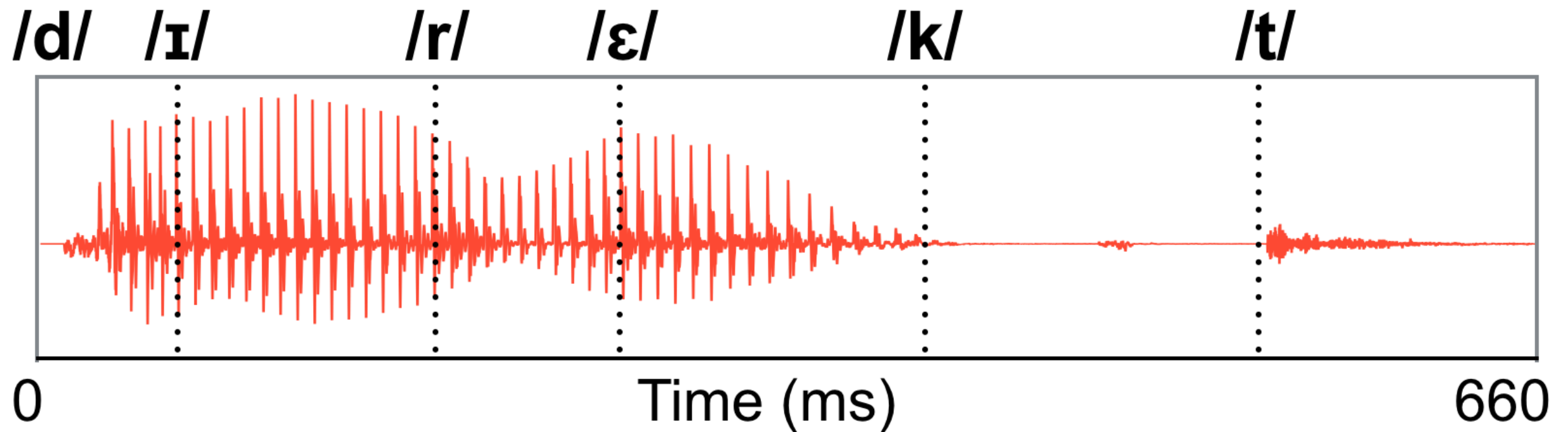


# Procedure & Analysis



# Procedure & Analysis

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# Model Setup

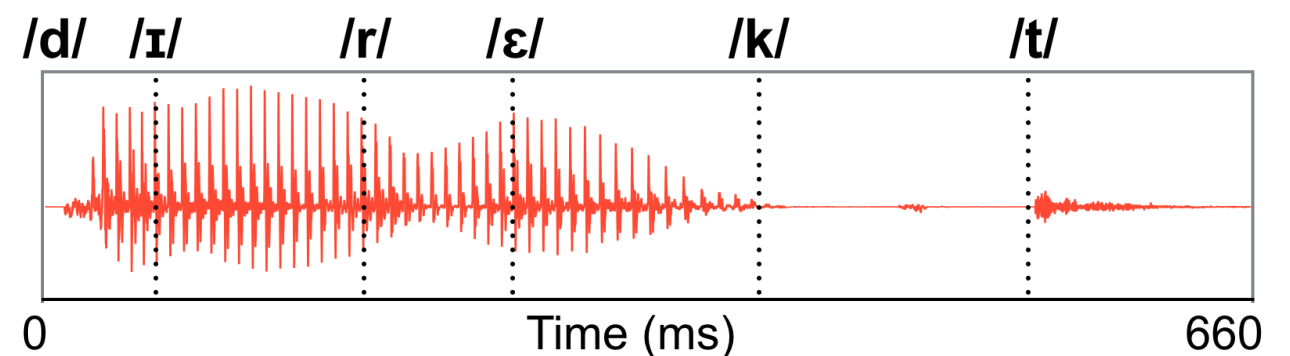
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- **Critical variables:**

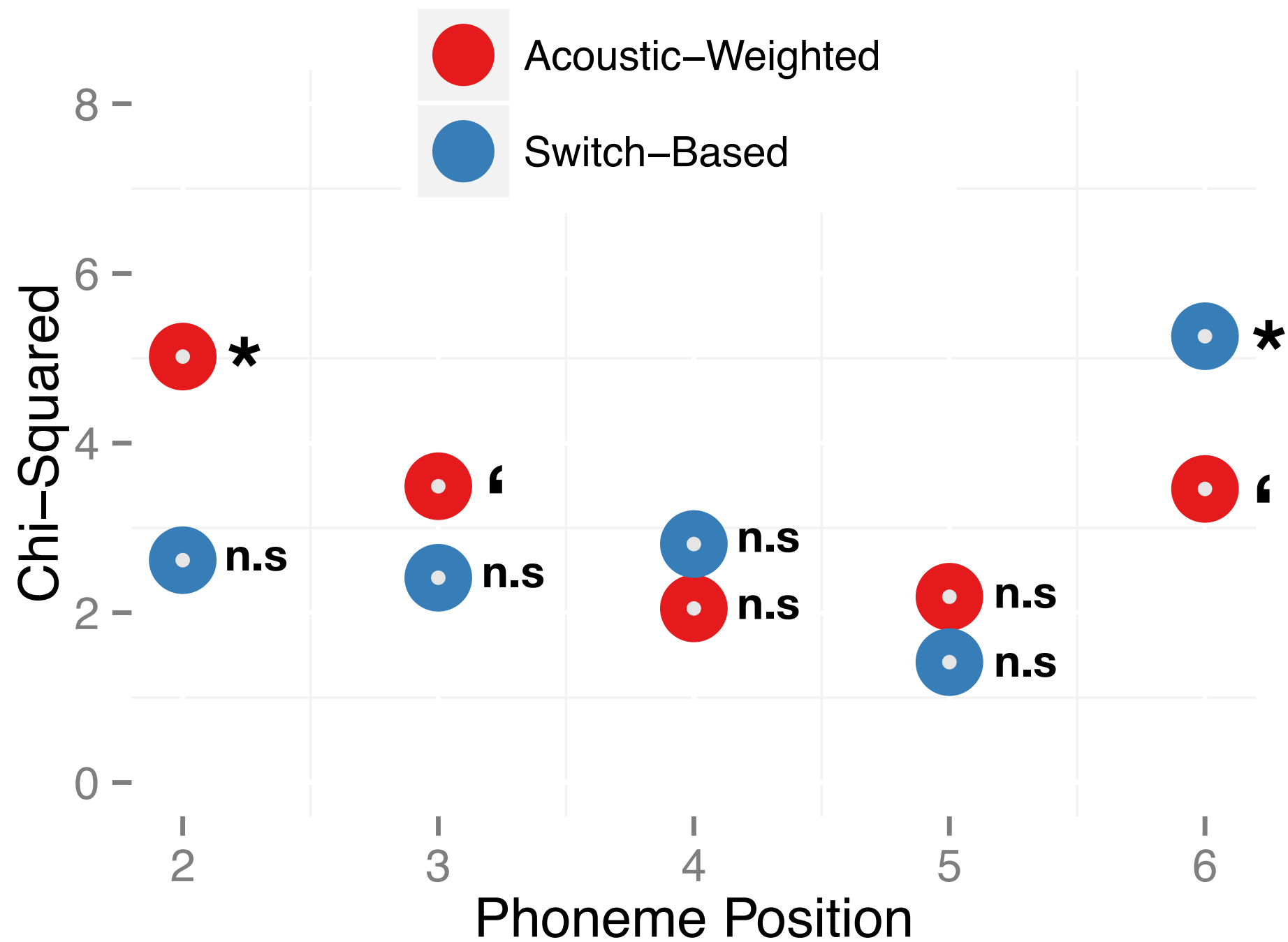
acoustic-weighted entropy  
acoustic-weighted surprisal  
switch-based entropy  
switch-based surprisal

- **Control variables:**

phoneme latency (ms)  
phoneme latency (number of phonemes)  
trial number  
block number  
stimulus amplitude  
phoneme pair  
ambiguity

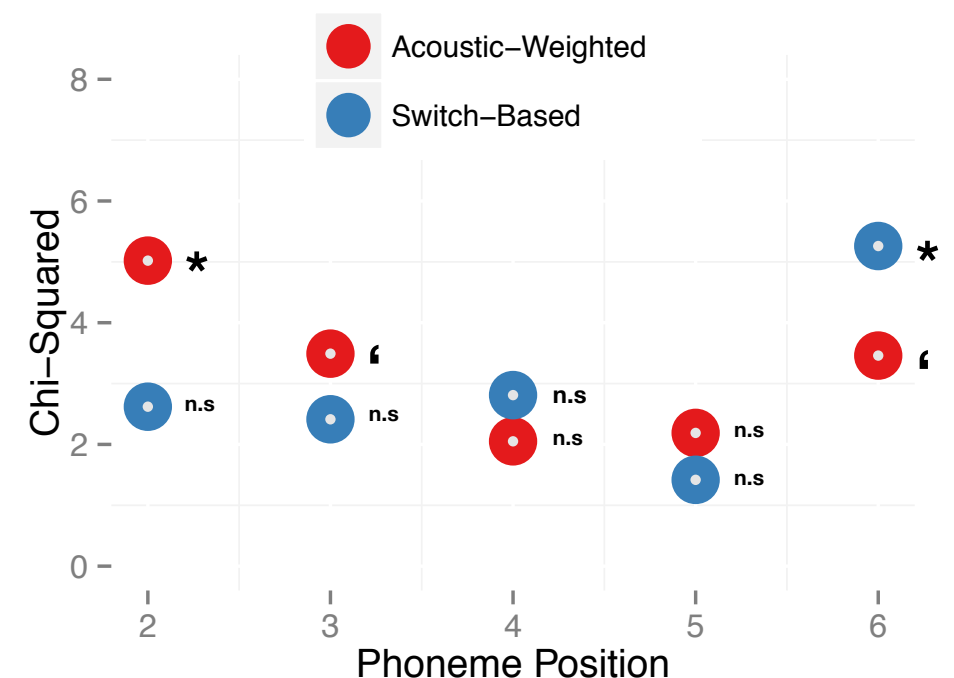


# Results



# Discussion

- Fine-grained acoustic information does weight lexical candidates
- There is a **dynamic interaction** between different levels of linguistic description: phonological <-> lexical
- Not a single heuristic applied in all situations: perhaps reflects that the **brain commits to an interpretation** of the phonological category after a certain period of time



# Research Answer

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Acoustic-phonetic uncertainty can  
weight activation at the lexical level



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- My supervisors, **Alec Marantz** and **David Poeppel**, as well as everyone in the **Neuroscience of Language Lab** and **Poeppel Lab**!



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🐦 [@GwilliamsL](https://twitter.com/GwilliamsL)

