

# The perceptual generalization of normalized cue distributions across speakers

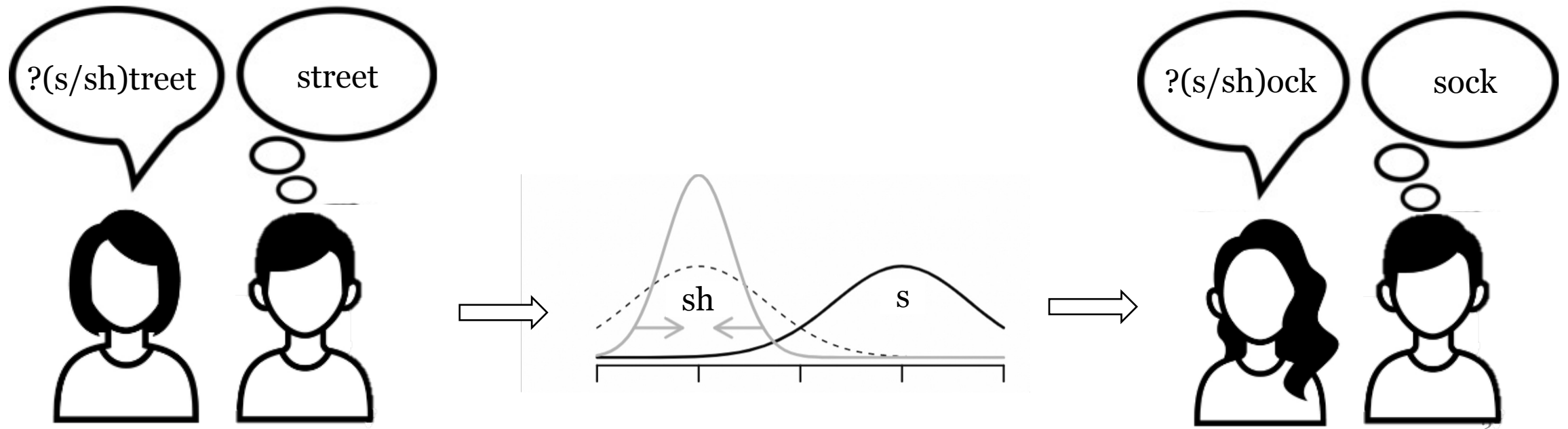
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# Perceptual learning and generalization

- Listeners make perceptual adjustments to adapt to talker-specific phonetic distributions. (Norris, McQueen, & Cutler, 2003)
- They also generalize the perceptual adjustments across different speakers. (Kraljic & Samuel, 2006; Reinisch & Holt, 2014; Xie et al., 2018).

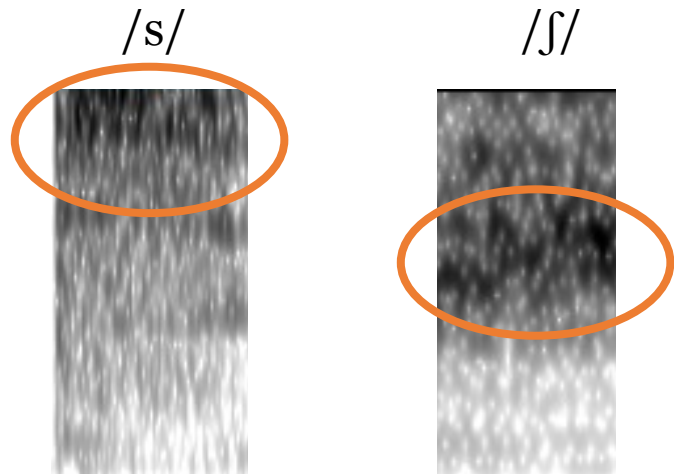


# Speech normalization

- Phonemic categorization is not only informed by raw phonetic distributions, but also *relative contextual cues* from the talker's speech.  
(e.g., Johnson, 1990, 2018; Port, 1979; Summerfield, 1975)

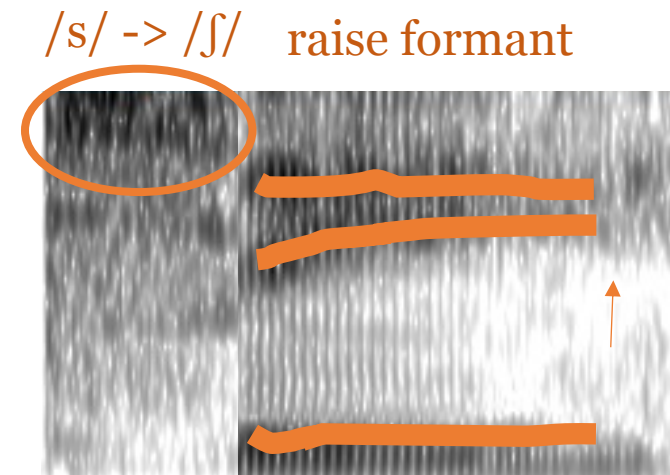
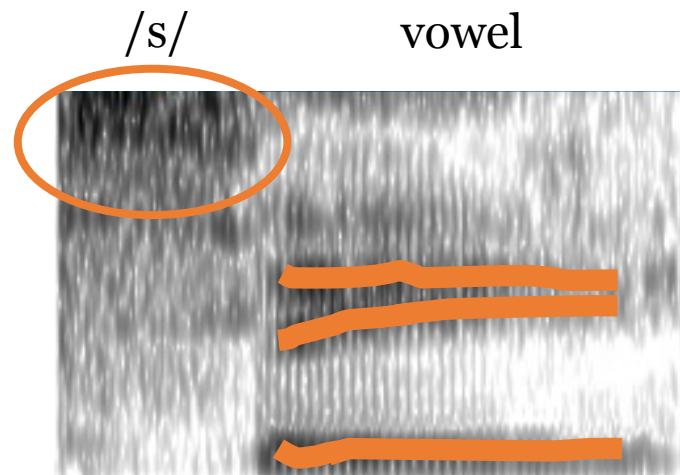
# Speech normalization of spectral cues

- The categorization of /s-ʃ/ varies with contextual vowel formants (Johnson, 1990, 2018)



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# Speech normalization of temporal cues

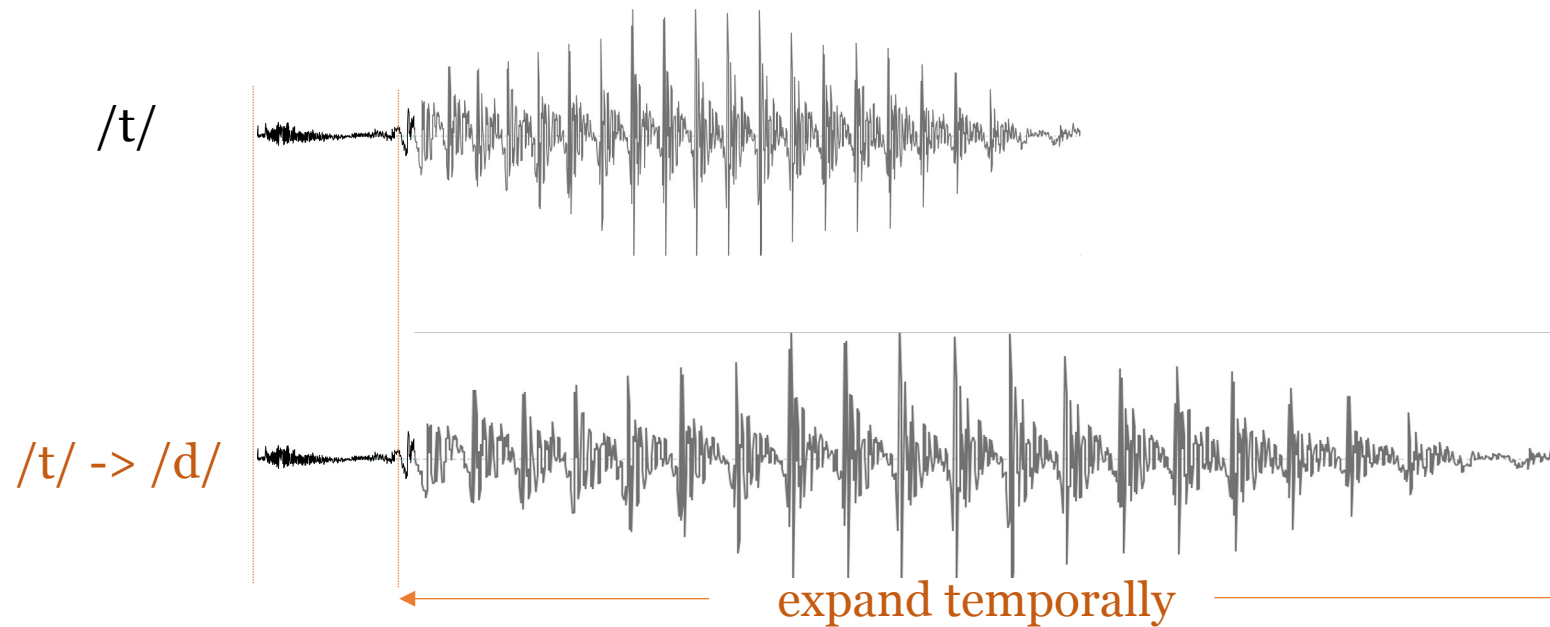
- The categorization of /t-d/ varies with contextual vowel duration  
(Summerfield, 1975; Port, 1979)

/t/ 

/d/ 

# Speech normalization of temporal cues

- The categorization of /t-d/ varies with contextual vowel duration  
(Summerfield, 1975; Port, 1979)



# Research Question

- In perceptual learning, do listeners learn and generalize raw phonetic cues or normalized cue distributions within a speaker's acoustic space?
  - Raw-distribution hypothesis
  - Normalized-distribution hypothesis
- The current study:
  - Experiment 1: spectral cues /s-f/
  - Experiment 2: temporal cues /t-d/



# Experiment 1: /s-f/

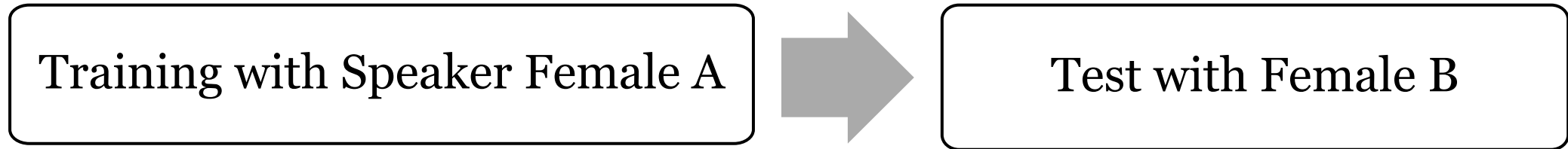
## Question:

- Would changing *contextual vowel formants* of a training speaker affect listeners' categorization of /s-f/ in a test speaker's speech?

## Subject:

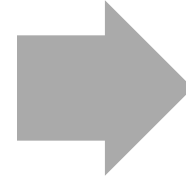
- 45 monolingual English speakers (20 men and 25 women) recruited through Prolific to participate online.
- Experiment implemented through PennController IBex.

# Experiment 1: Method



# Experiment 1: Method

Training with Speaker Female A



Test with Female B

- 51 trials of spoken word identification

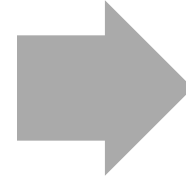


rehearsal

reversal

# Experiment 1: Method

Training with Speaker Female A

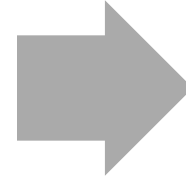


Test with Female B

- 51 trials of spoken word identification
- - 17 words containing /s/
- -17 words containing /ʃ/
- -17 fillers with no /s ʃ/

# Experiment 1: Method

Training with Speaker Female A



Test with Female B

- 51 trials of word identification

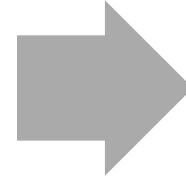


sake

shake

# Experiment 1: Method

Training with Speaker Female A



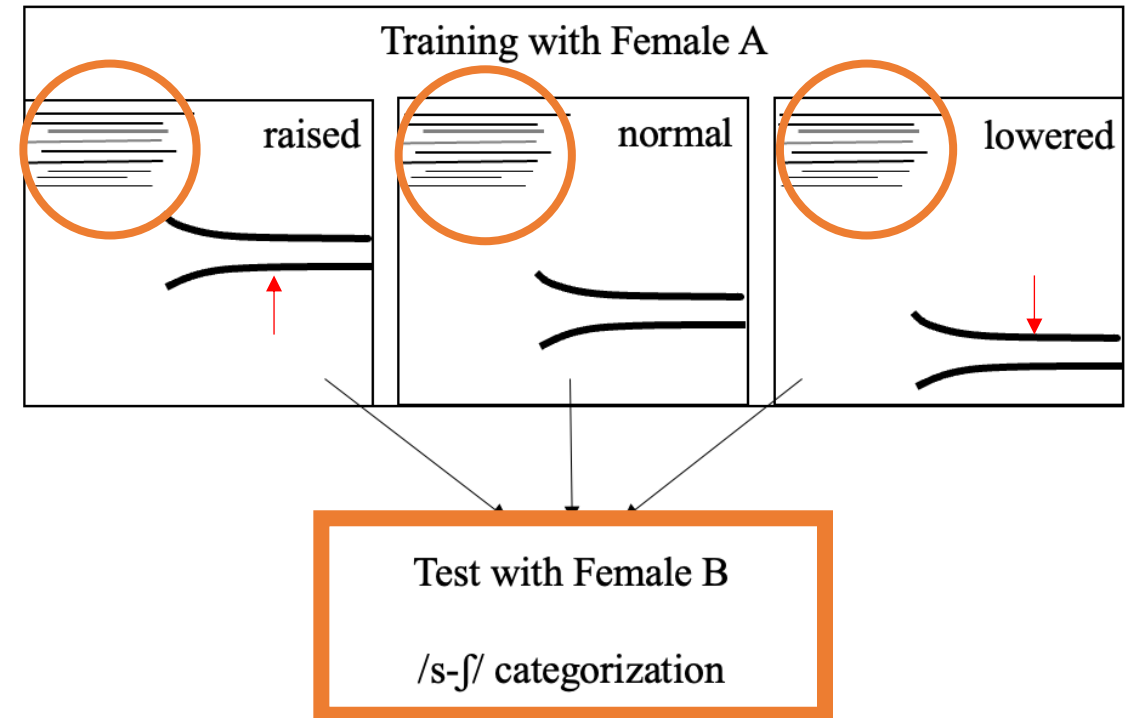
Test with Female B

- 51 trials of word identification
- - 35 /s ʃ/ minimal pairs
  - 5 steps x 7 words
  - *same, sign, seat, shelf, shake, shell, shy*
- - 16 filler trials with no /s ʃ/

# Experiment 1: Method

Participants assigned to 3 experiment conditions (N=15 on each condition):

- identical test phase
- identical /s f/ in the training stimuli
- *different* context vowel formants of the training stimuli:
  - Normal: unaltered
  - Raised: scale formants by 1.2
  - Lowered: scale formants by 0.8



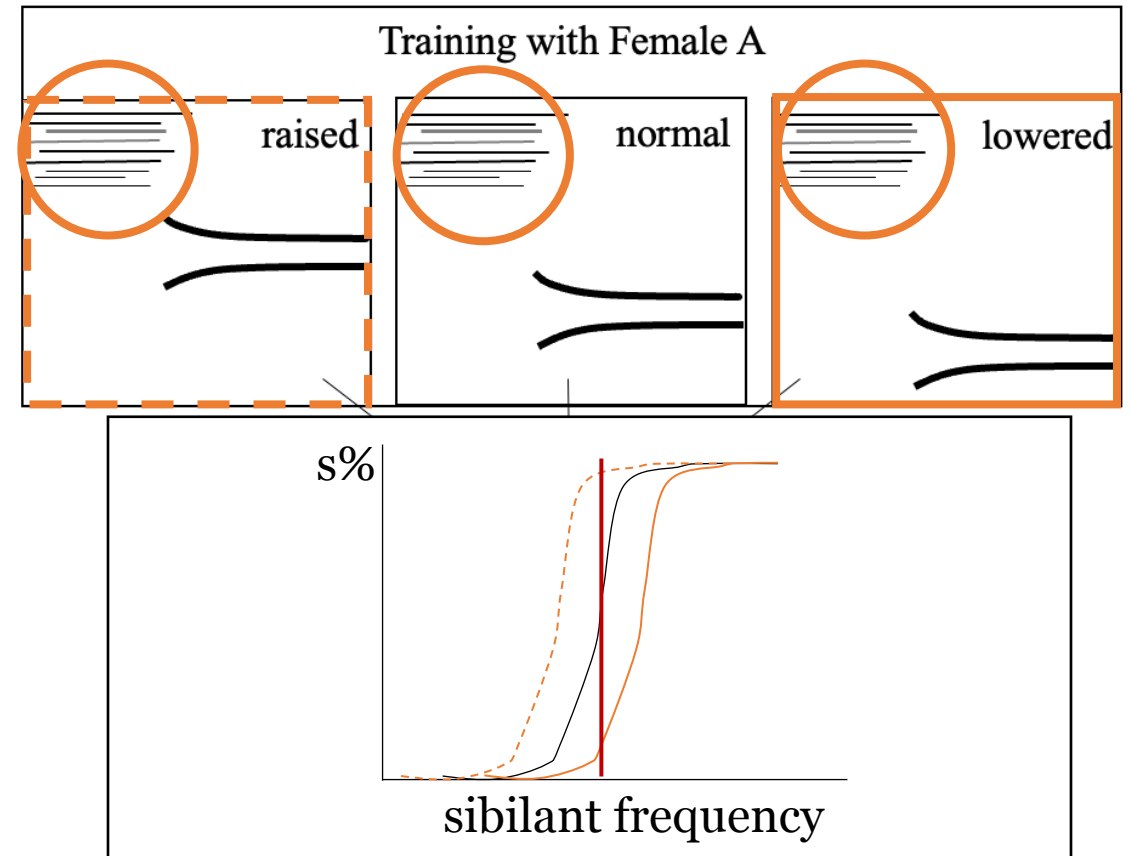
# Experiment 1: Hypotheses

Raw-distribution hypothesis:

- Predicts that participants across conditions do not differ

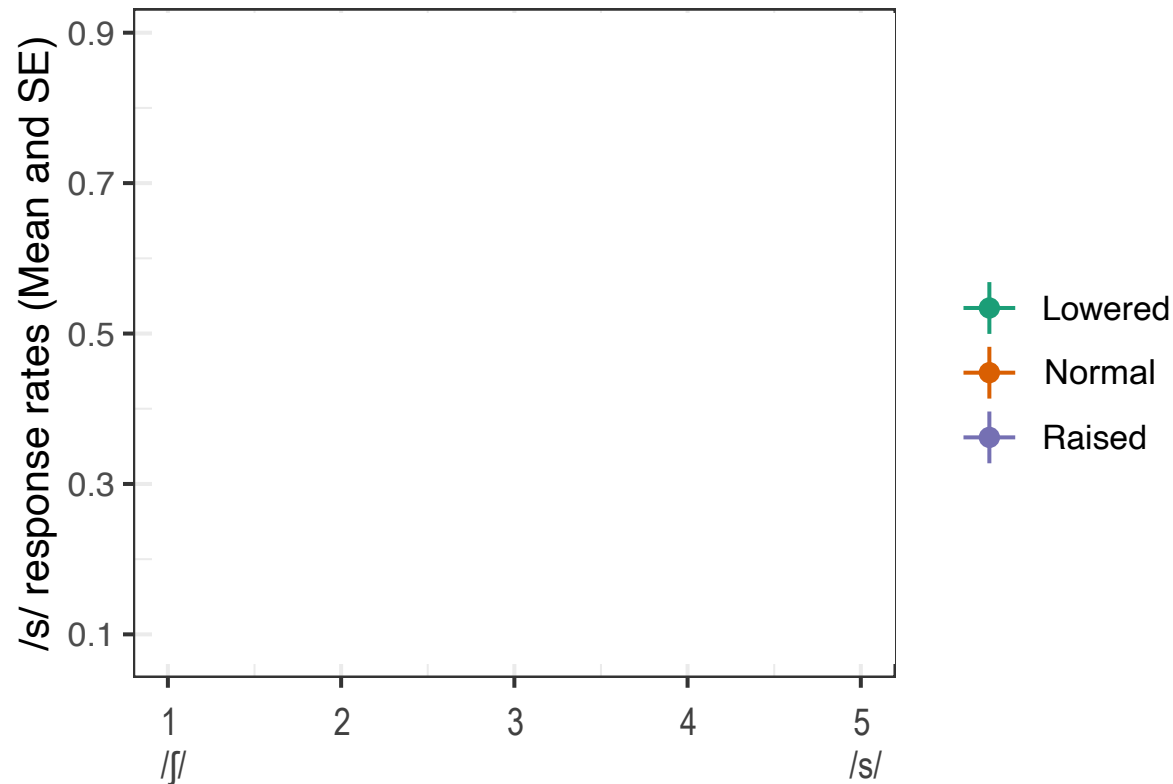
Normalized-distribution hypothesis :

- The proportion of /s/: raised > normal > lowered





# Experiment 1: Results



- /s/ response rate: raised > normal > lowered
- Normalized distribution hypothesis 😊
- $response \sim step * condition + (step|subject) + (step|word)$ 
  - Step:  $\beta = 1.82, p < 0.001$
  - Condition (raised-lowered):  $\beta = 1.54, p = 0.02$

# Intermediate summary

- In the perceptual generalization of sibilants across speakers, changing contextual spectral cues of a training speaker would affect listeners' sibilant categorization of a test speaker
- The perceptual learning of *spectral* cues involves some degree of knowledge and computation about speaker-normalized distributions.
- Will the pattern hold for *temporal* cues?

# Experiment 2: /t-d/

## Question:

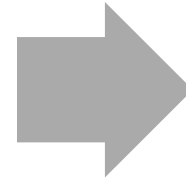
- Would changing **contextual temporal** cues of a training speaker affect listeners' categorization of /t-d/ in a test speaker's speech.

## Subject:

- 45 English monolinguals (23 men and 22 women) recruited through Prolific to participate in the experiment online.
- Experiment implemented through PennController IBex.

# Experiment 2: Method

Training with Speaker Female A



Test with Female B

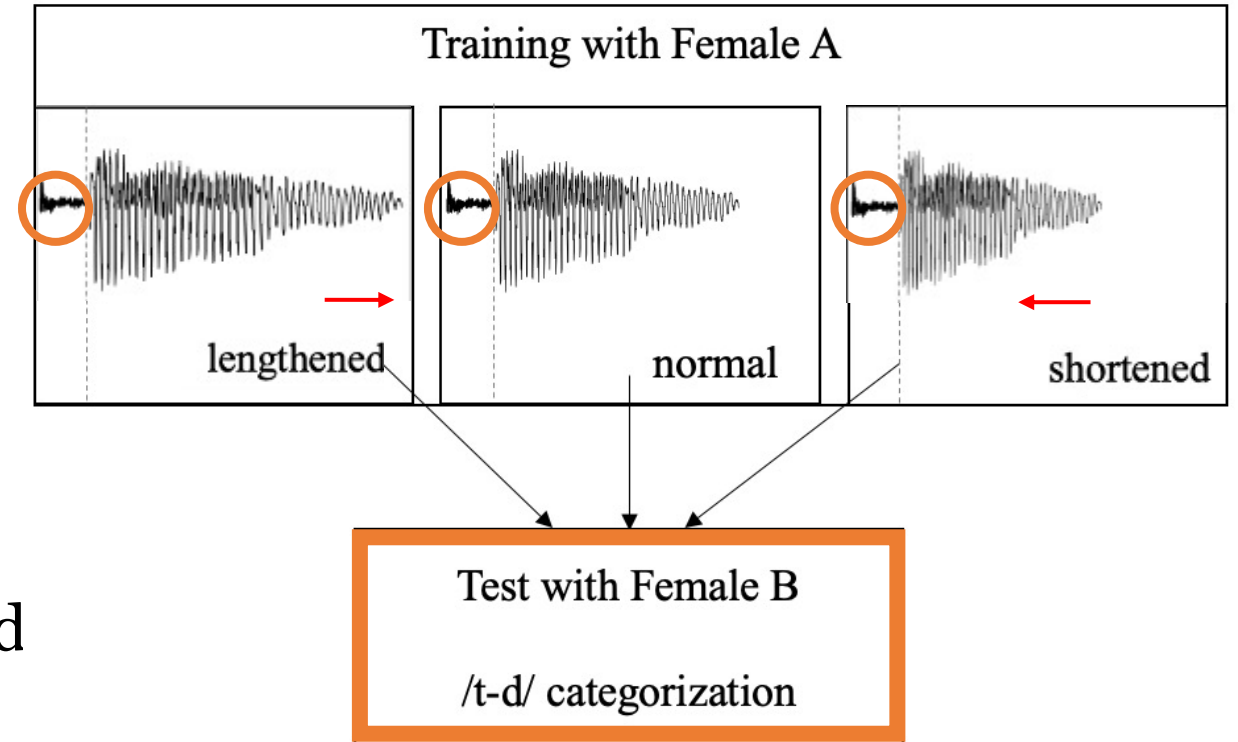
- 51 trials of spoken word identification
- - 17 words containing /t/
- -17 words containing /d/
- -17 fillers with no /t d/

- 51 trials of word identification
- - 35 /t d/ minimal pairs
  - 5 steps x 7 words
  - *tear, tie, town, touch, time, tip, toes*
- - 16 filler trials with no /t d/

# Experiment 2: Method

Participants assigned to 3 experiment conditions (N=15 on each condition):

- identical test phase
- identical /t d/ in the training stimuli
- *different* context speech rates of the training stimuli
  - Normal: unaltered
  - Lengthened: temporally expanded by 1.7
  - Shortened: temporally compressed by 0.7



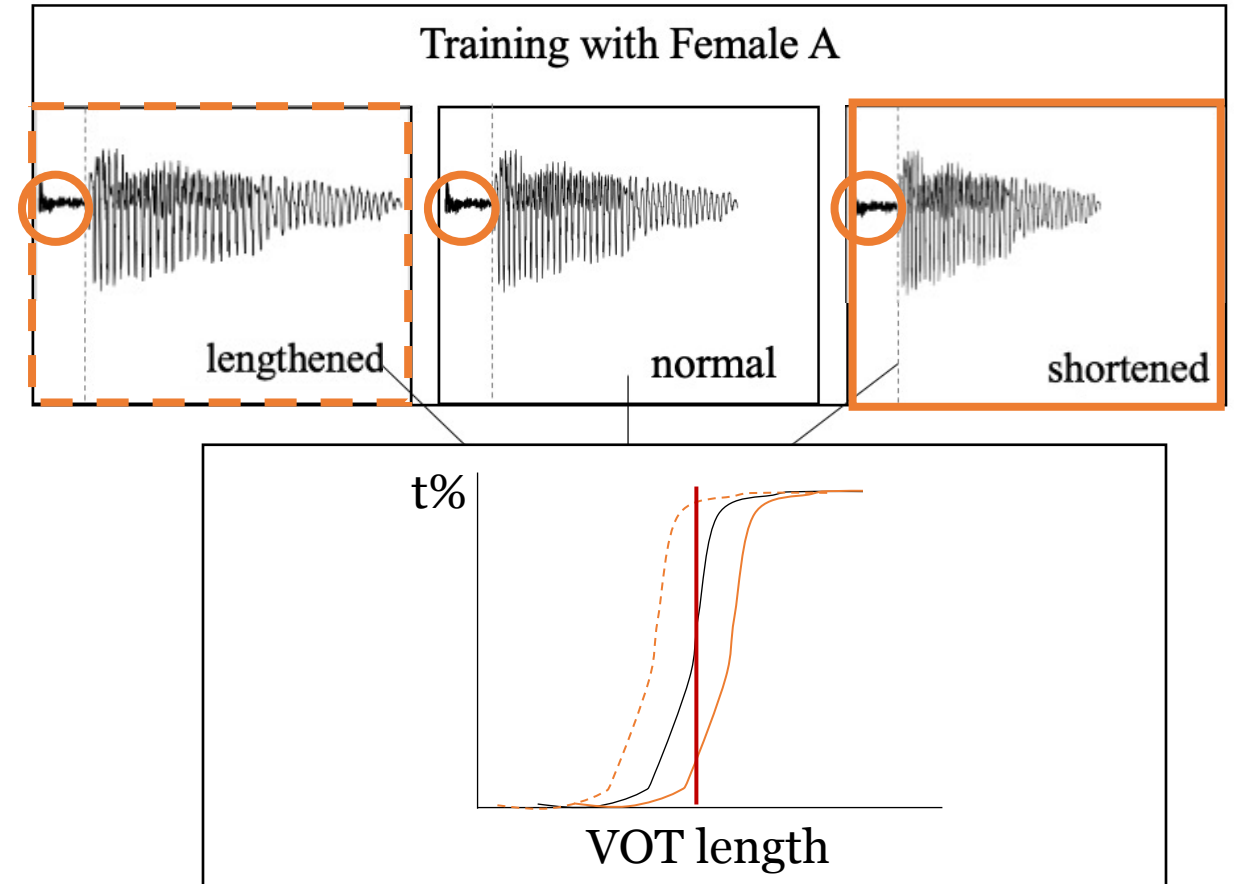
# Experiment 2: Hypotheses

Raw-distribution hypothesis:

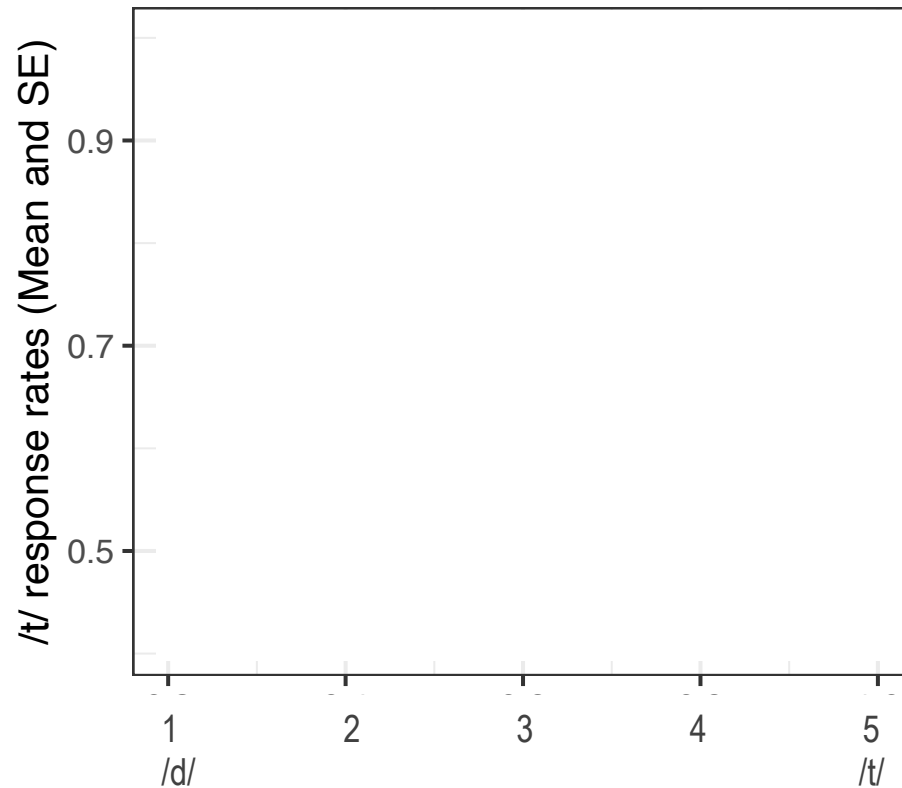
- Predicts that participants across conditions do not differ

Normalized-distribution hypothesis :

- The proportion of /t/: lengthened > normal > shortened



# Experiment 2: Results



- /t/ response rate: lengthened > normal > shortened
- Normalized distribution hypothesis 😊
- $response \sim step * condition + (step | subject) + (step | word)$ 
  - Step:  $\beta = 1.18, p < 0.001$
  - Step: Condition (lengthened-shortened):  $\beta = 0.54, p = 0.01$

# Discussion

- Studies on perceptual learning and speech normalization were usually discussed in the lines of different theoretical frameworks (e.g., abstractionist vs. exemplar theory).
- The study provides preliminary evidence of their interaction, i.e., listeners learn and generalize speaker-normalized distributions.
- Our findings shed lights on the possibility of incorporating speech normalization mechanisms into current perceptual learning models (e.g., Kleinschmidt & Jaeger, 2015)



# Thank you!

- Reach me at *wei.lai@vanderbilt.edu*!