

The Role of Source and Filter Characteristics in Human Talker Identification: Experiments with Laryngeal and Electrolarynx Speech

Tyler K. Perrachione¹, Cara E. Stepp^{2,3,5}, Robert E. Hillman^{2,3,4}, Patrick C.M. Wong^{6,7,8}

¹Department of Brain & Cognitive Sciences, Massachusetts Institute of Technology, ²Harvard-MIT Division of Health Science & Technology, ³Massachusetts General Hospital Center for Laryngeal Surgery & Voice Rehabilitation, ⁴Department of Surgery, Harvard Medical School, ⁵Departments of Computer Science and Engineering & Rehabilitation Medicine, University of Washington, ⁶Roxelyn & Richard Pepper Department of Communication Sciences & Disorders, Northwestern University, ⁷Northwestern Interdepartmental Neuroscience Program, Northwestern University, ⁸Department of Otolaryngology – Head and Neck Surgery, Northwestern University

Abstract

Differences in individuals' vocal anatomy and physiology result in unique acoustic features of their vocalizations. Humans are exceptionally attuned to these variations and use them to identify familiar individuals. Although these abilities are often called "voice recognition", talker identity cues actually arise through interactions between acoustic excitation produced at the source (typically, the larynx) and both static and dynamic properties of the filter (vocal tract, articulators, and their manipulations during speech). We investigated the differential contributions of source- and filter-related information to talker identification through four experiments using laryngeal (typical) and electrolarynx speech from 5 talkers. Using an electrolarynx energy source removed individual differences in vocal anatomy, leaving only unique filter properties for talker identification.

Listeners learned talker identity best from typical, laryngeal speech, which contained both unique source and filter cues. Listeners were also able to learn talker identity from electrolarynx speech, which homogenized talker source characteristics. Curiously, listeners did not generalize talker identity across source mechanisms: Training on laryngeal or electrolarynx speech resulted in chance performance identifying the same talkers using the other source mechanism. We consider the implications of these results for models of talker identification and articulatory compensation during electrolarynx use.

Background

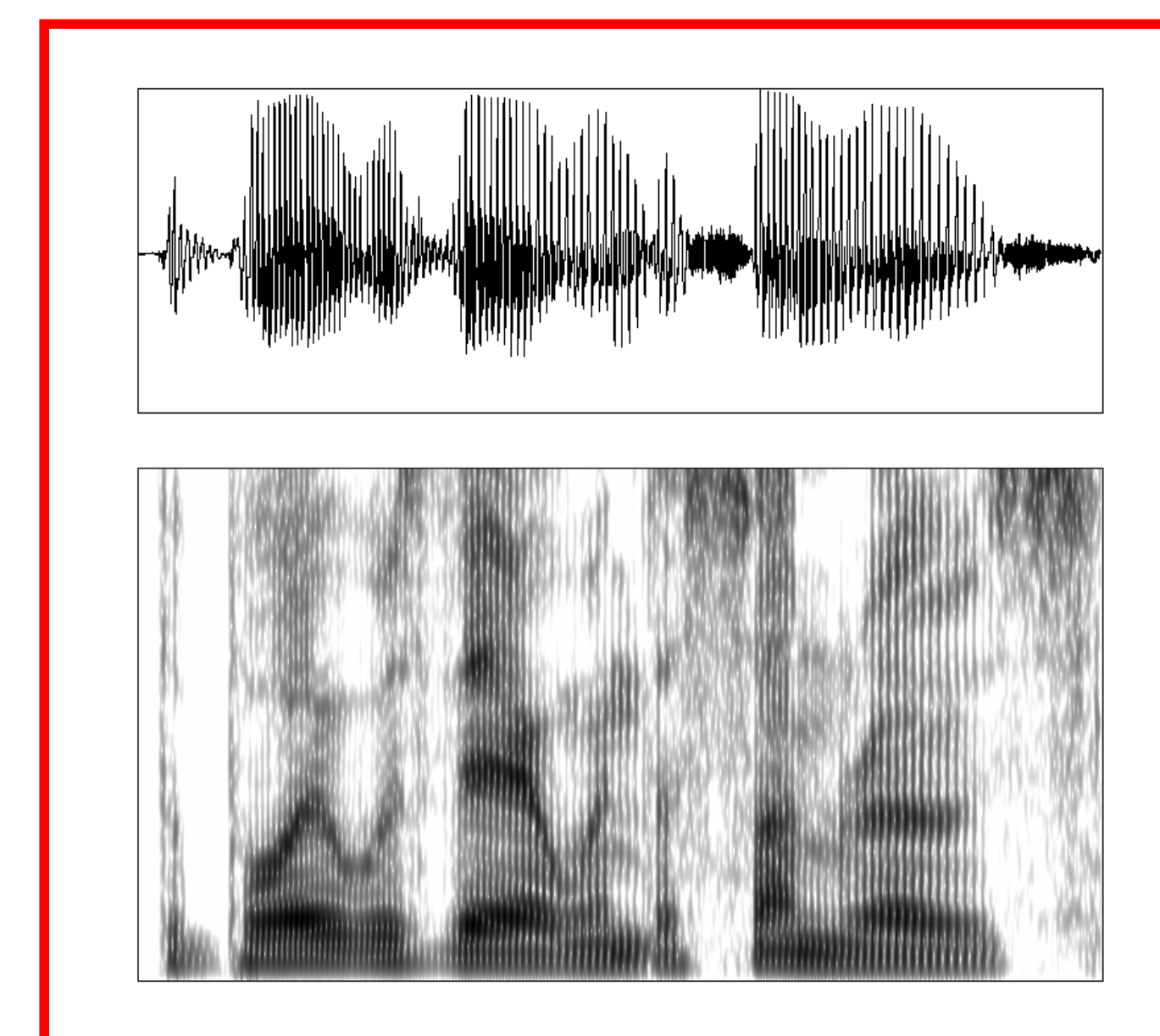
Talker identity is the product of interacting acoustic cues:

- **Source characteristics** of the voicing mechanism
 - Laryngeal anatomy and physiology
 - F_0 and its dynamics, glottal waveform, etc.
- **Filter characteristics** of the vocal tract
 - Anatomy and physiology of the pharyngeal, oral, and nasal cavities
 - Constrain the range of resonance ($F1$, $F2$, etc.)
- **Dynamic manipulation** of articulators
 - Socioculturally acquired phonetic features
 - Variation due to language, dialect, idiolect

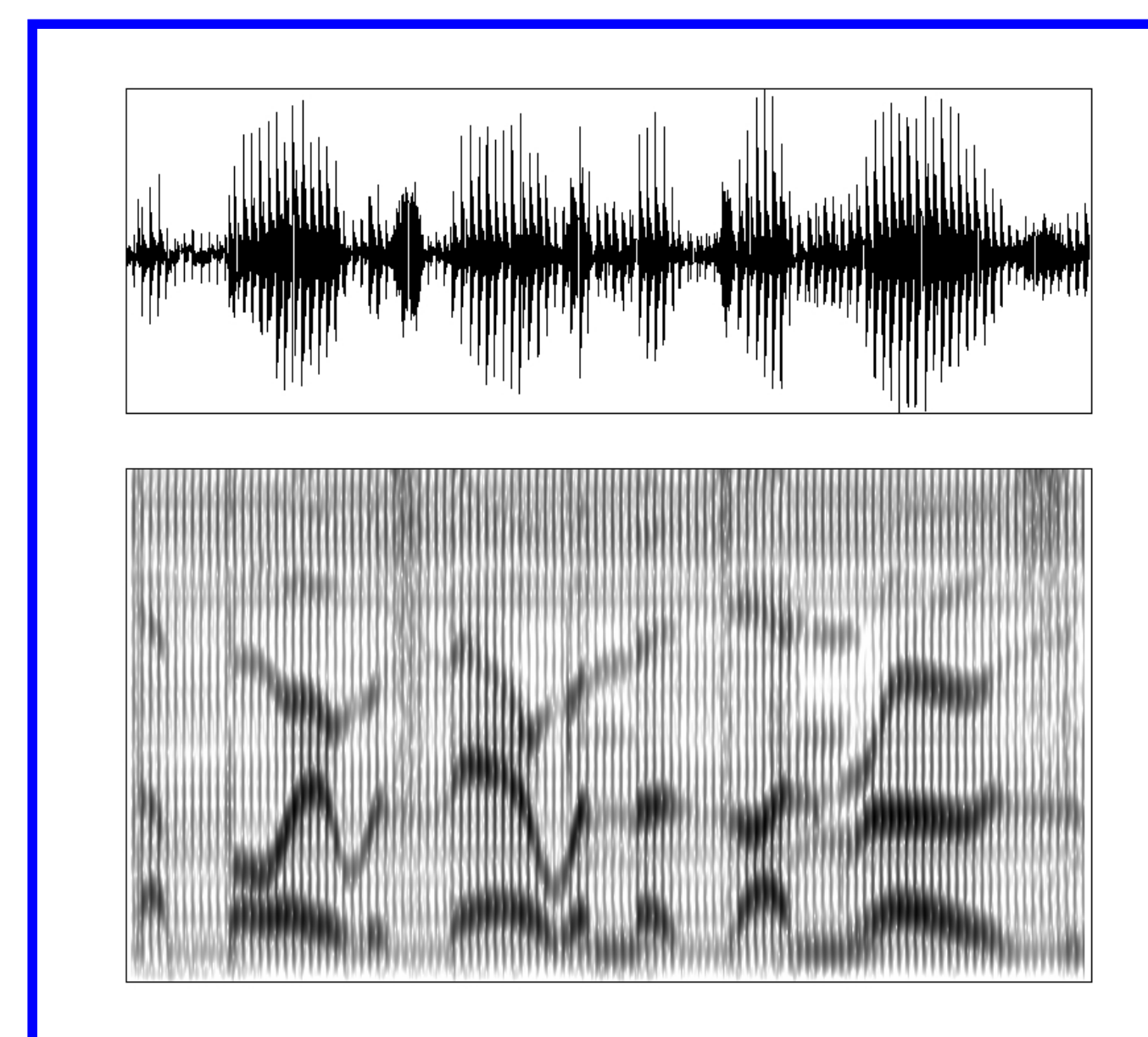
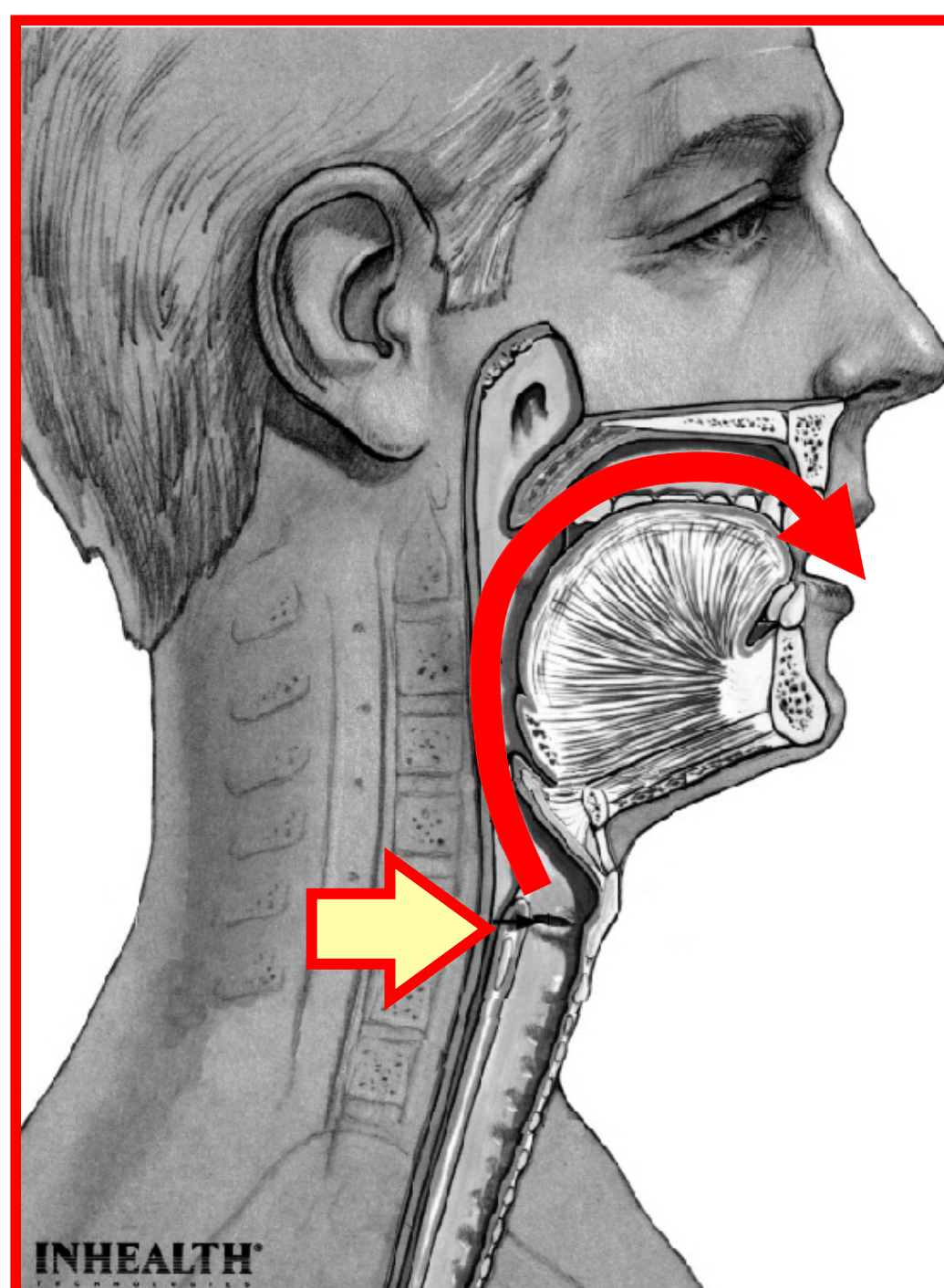
Carrell, (1984); Perrachione & Wong (2007);

How do source and filter characteristics differentially contribute to perception of talker identity?

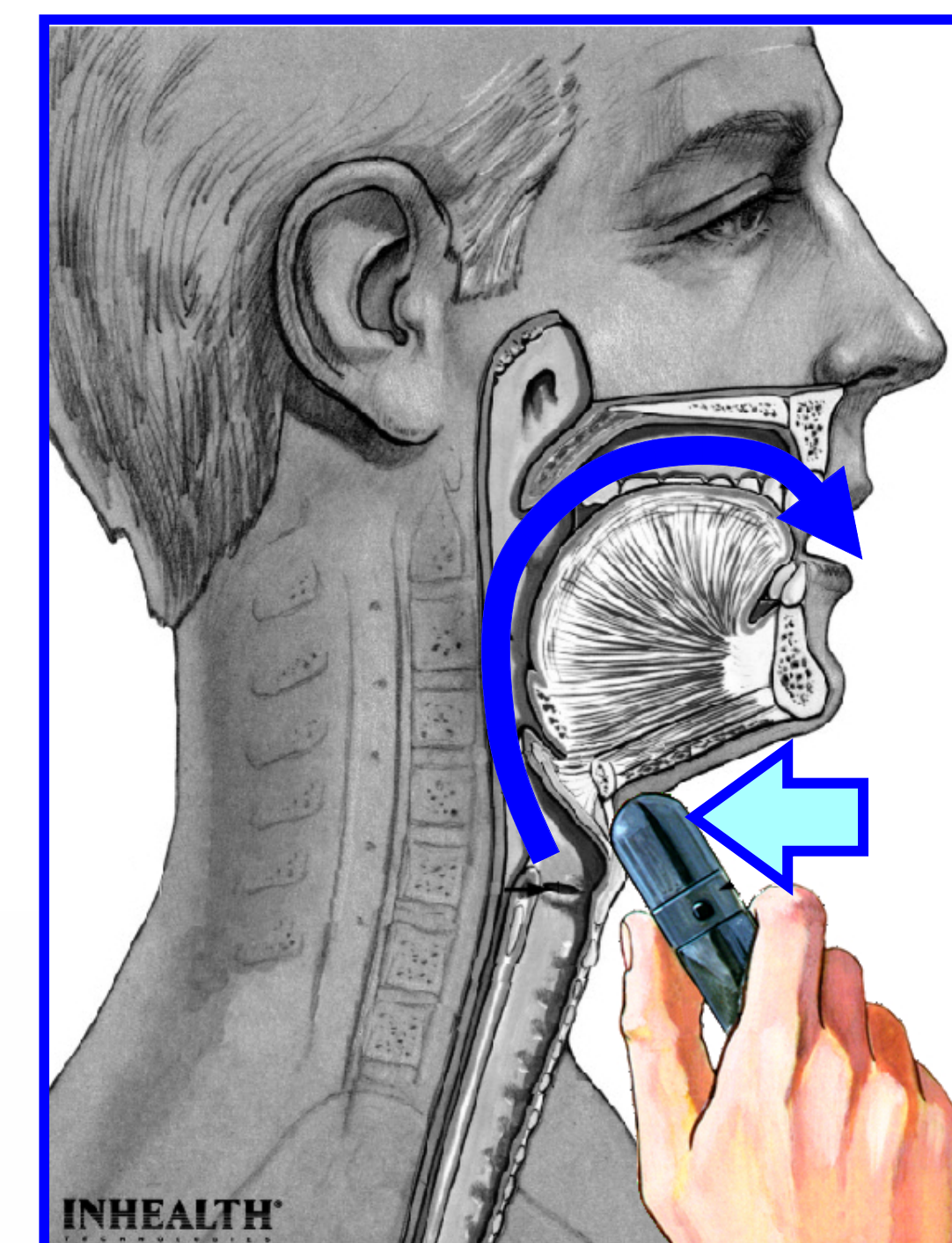
- **Experimental challenge:** how to separate the 1-to-1 correspondence between unique vocal sources and filter characteristics (i.e. within a single talker)
- **Electrolarynx:** a battery-powered device that provides a mechanical voice source through the tissues of the neck
 - Replacement voice source for total laryngectomy patients
 - **Homogenizes source characteristics while preserving individual differences in filter characteristics**
 - e.g., Listeners can still distinguish male/female EL users based on formant spacing (Brown & Feinstein, 1997)



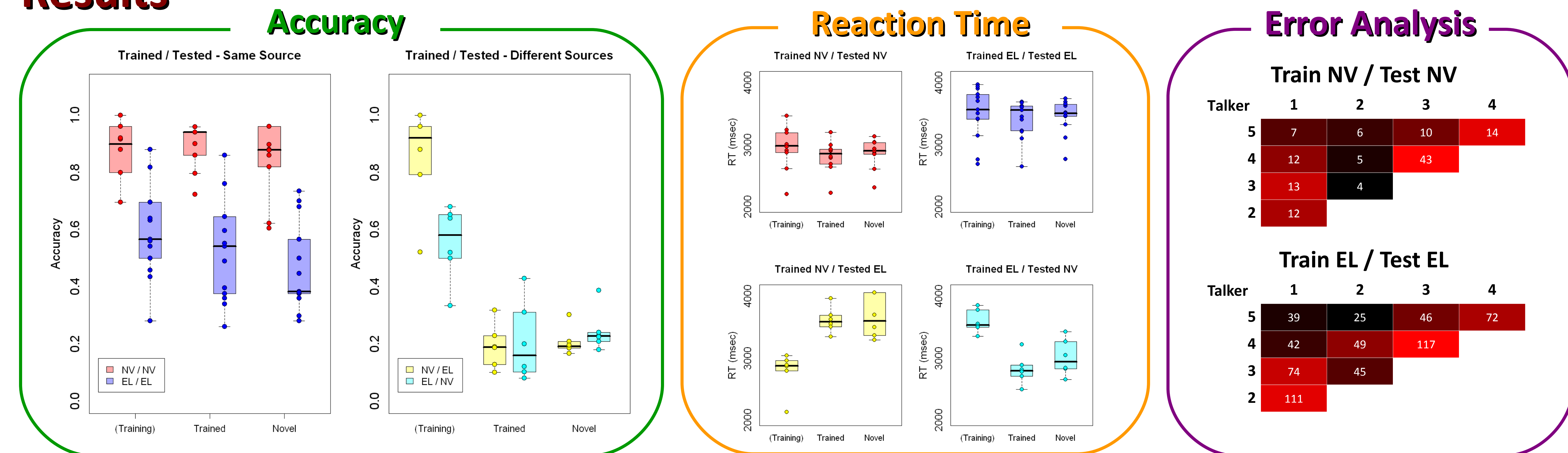
Laryngeal Speech



Electrolarynx Speech



Results



Source Mechanism: “(Training)”

- Better talker identification from NV training
 - $t(33) = 5.875, p < 1.4 \times 10^{-6}$
- Faster reaction time to NV stimuli
 - $t(33) = -5.444, p < 5 \times 10^{-6}$
 - Evitts & Searle (2006)

Sentence Content: “Trained” vs. “Novel”

- More accurate ID from trained sentences
 - $F(1,31) = 10.250, p < 0.0032$
- Faster RT to trained sentences
 - $F(1,31) = 4711.653, p < 2 \times 10^{-35}$
- No interaction with source mechanism.

Filter-Characteristics Generalization:

- Listeners accuracy was at chance for testing on the untrained source mechanism:
 - NV-EL: $t(5) = -0.009, p = 0.993$
 - EL-NV: $t(5) = 0.644, p = 0.548$

Patterns of Correct Identification and Errors:

- Across source mechanisms, listeners differed in the talkers they found most identifiable:
 - $\chi^2(4) = 16.703, p < 0.0025$
- Patterns of errors across source mechanism showed some similarities, but differed significantly overall
 - $\chi^2(9) = 27.875, p < 0.001$

Methods

Recordings

- 15 male native American English-speakers with no discernable accent
 - Ages 20-38 years, mean = 26.6 years
- 14 sentences (IEEE, 1969) recorded at 50 kHz
 - **Normal laryngeal voice (NV)** and using an **electrolarynx (EL)**
 - TrueTone™ electrolarynx (Griffin Labs), fixed F_0 of 109Hz

Intelligibility Assessment & Stimulus Selection

- 8 native English-speaking listeners judged pairs of EL recordings
 - “Which recording is more intelligible?”
- 210 stimulus pairs (15 talkers $\times$ 14 sentences)
 - Recordings in a pair were of the same sentence
 - Each talker was paired against every other talker equally
- Intelligibility rankings determined following Meltzner & Hillman (2005)
 - Most intelligible talker used as an example stimulus
 - Next 5 most intelligible talkers used as stimuli for identification

Talker Identification Training & Testing

Conditions: Parametric training-testing paradigms investigated generalization of talker-identification abilities across source mechanisms

- Train **NV** — Test **NV** (N = 10)
- Train **EL** — Test **EL** (N = 13)
- Train **NV** — Test **EL** (N = 6)
- Train **EL** — Test **NV** (N = 6)

Subjects: Undergraduate students, native speakers of American English, normal speech and hearing, N = 35

Paradigm:

- **Training:** Learn to identify (with feedback) 5 talkers from 5 training sentences with training source mechanism (NV or EL)
- **Generalization & Post-test:** Identify those talkers from both trained and novel sentences with testing source mechanism (NV or EL)

Conclusions:

- Individuals are uniquely identifiable even from EL speech.
- Listeners' perceptual space for talkers differs across source mechanisms (based on identification errors)
- Likewise, Training talker identification on one source mechanism does not generalize to speech from the other mechanism.
 - Talker identity may emerge as a “gestalt” from a number of underlying cues; is more than the sum of its acoustic parts.
 - Talkers may employ different articulatory strategies to increase intelligibility during naïve use of an EL device.

References

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Contact

Tyler K. Perrachione tkp@mit.edu
<http://web.mit.edu/tkp/www/>

Cara E. Stepp cstepp@alum.mit.edu
<http://faculty.washington.edu/cstepp/>