

Link to GitHub

Introducing LingTube: An open-source toolkit for linguistic analysis of YouTube data

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Key Points

- **LingTube** is a suite of tools for automating the downloading and processing of captioned YouTube audio for textual and/or phonetic analysis.
- We present ongoing exploratory work using LingTube to investigate phonetic features in the English of Asian North American (ANA) speakers.

YouTube Background

Vast, yet relatively untapped source of publicly-available linguistic data (Schneider, 2016)

- **Advantages:**
 - ‘Naturalistic’, (auto-)captioned speech data representing various contexts
 - Large user base → improved access to lesser-studied language varieties/communities
- **Disadvantages:**
 - Not straightforward to download and process raw data
 - Manual work needed to screen usable videos/speech (BG music, noise, audio quality, etc.)
 - Minimal, inferred or unavailable speaker background information

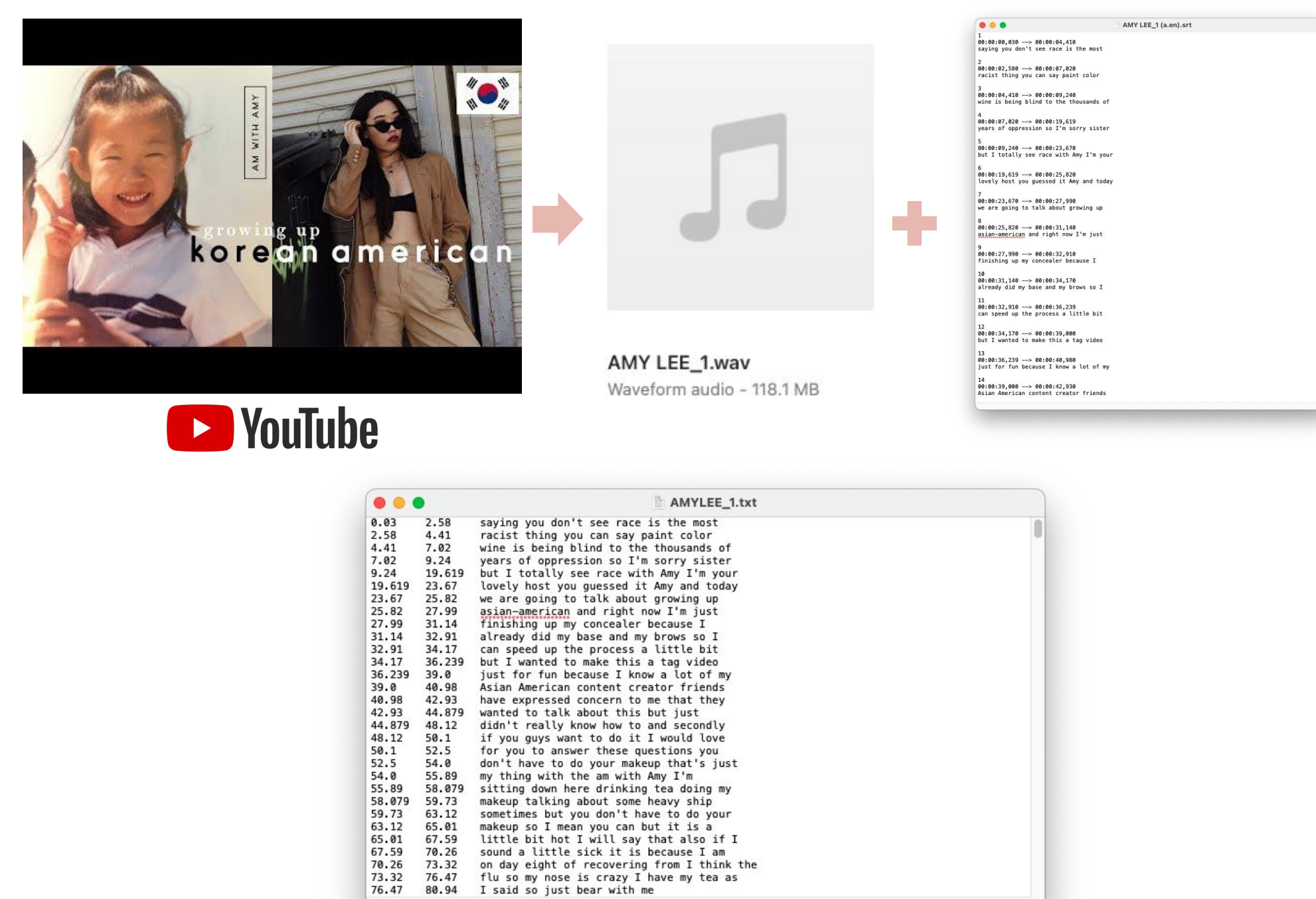
Some previous uses of YouTube:

- Lee, 2017: Style-shifting of one speaker across contexts
- Coats, 2020: Articulation rate across US regions
- Kramer, 2021: Cross-linguistic patterns of dependency length minimization

LingTube Pipeline: <https://github.com/Narquelion/LingTube>

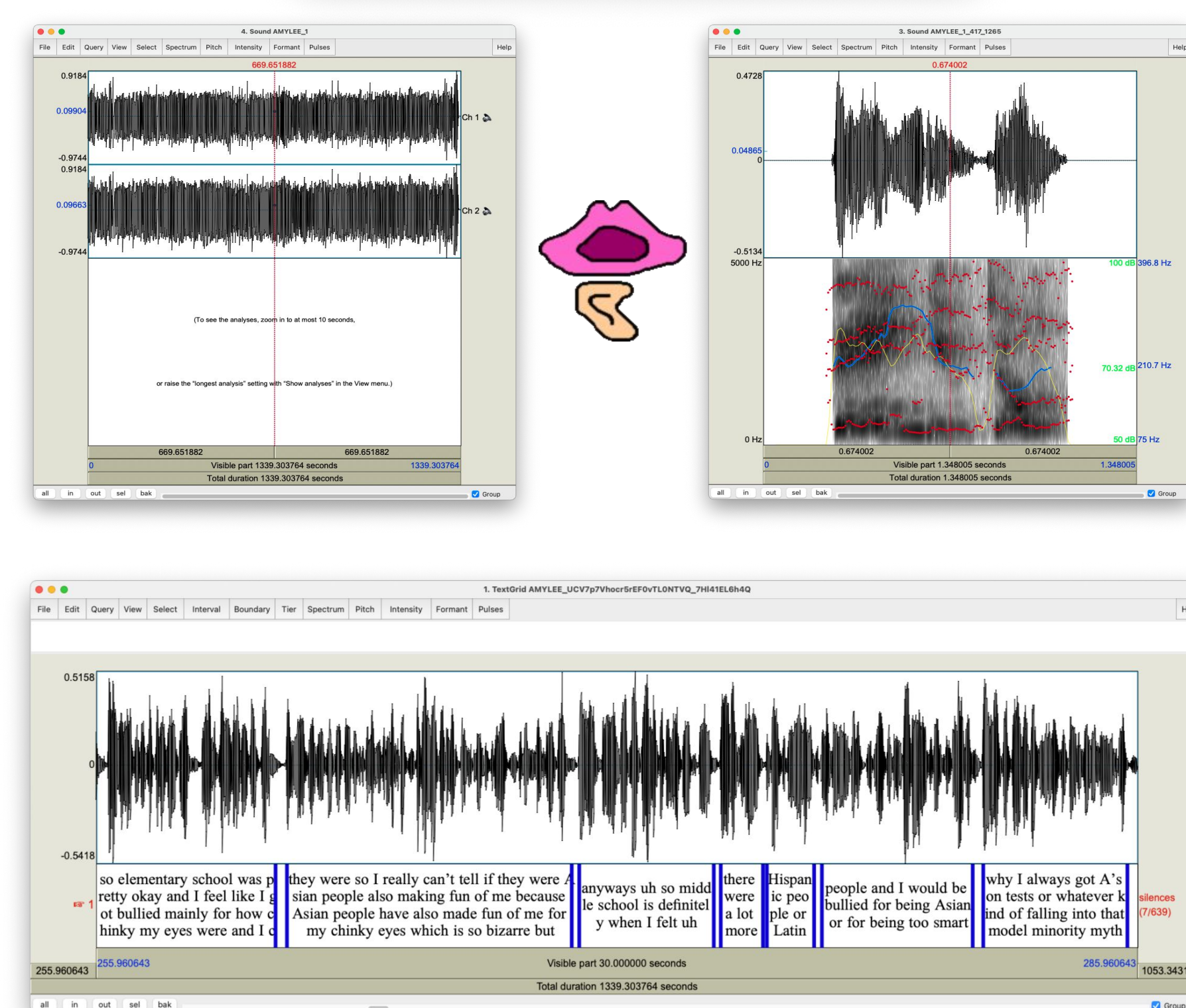
Base: Download & pre-process captions

1. **Scrape channels** Scrape video URLs and about page information from channel(s)
2. **Scrape videos** Download video audio and captions
3. **Clean captions** Convert SRT files to tidy TXT files with columns; automatic cleaning of common issues
4. **Correct captions** Open caption text file, YouTube video, and GUI for user to manually correct captions



YouSpeak: Process audio and transcript for forced alignment

1. **Convert audio** Converts MP4 files to mono WAV
2. **Chunk audio** Create textgrid with utterances separated by breath breaks and saves audio chunks
3. **Validate chunks** Allow user to listen to audio chunks, validate transcript, and mark as usable/not
4. **Create textgrids** Add validated text to textgrid; optionally copy files to a folder for forced alignment



YouTube ANA English Analysis

The corpus

- 66 YouTubers: 46 ANA-identified, 3 multiracial ANA, 17 non-ANA
 - All presented as women and were East or Southeast Asian
- Used video(s) discussing (ANA) identity; otherwise, personal Q&A or conversational speech
- Demographic/language info coded from videos/“About” page

Steps

1. Identified video URLs via “Growing Up Asian American tag”, etc.
2. Scraped channel info, video captions, and audio
3. Cleaned captions; RAs manually corrected auto-transcription
4. Converted audio to mono WAV and chunked into utterances
5. RAs coded audio usability and validated transcript alignment
6. Created TextGrids with time-aligned transcribed utterances
7. Forced-aligned using Montreal Forced Aligner (McAuliffe et al., 2017)
8. Extracted vowel duration and formants (25%, midpoint, 75%)

Analysis

- Assessed phonetic features linked to ANA identity (Bauman, 2016)
 - Back-vowel retraction
 - Monophthongization
 - “Syllable-timed rhythm”
- Performed hierarchical clustering using Ward’s method on 5 measures per speaker
 - Two clusters but no pattern based on (non-)ANA identification
- Fig 1 shows that ANA vs. non-ANA speakers are distributed across each measure’s range

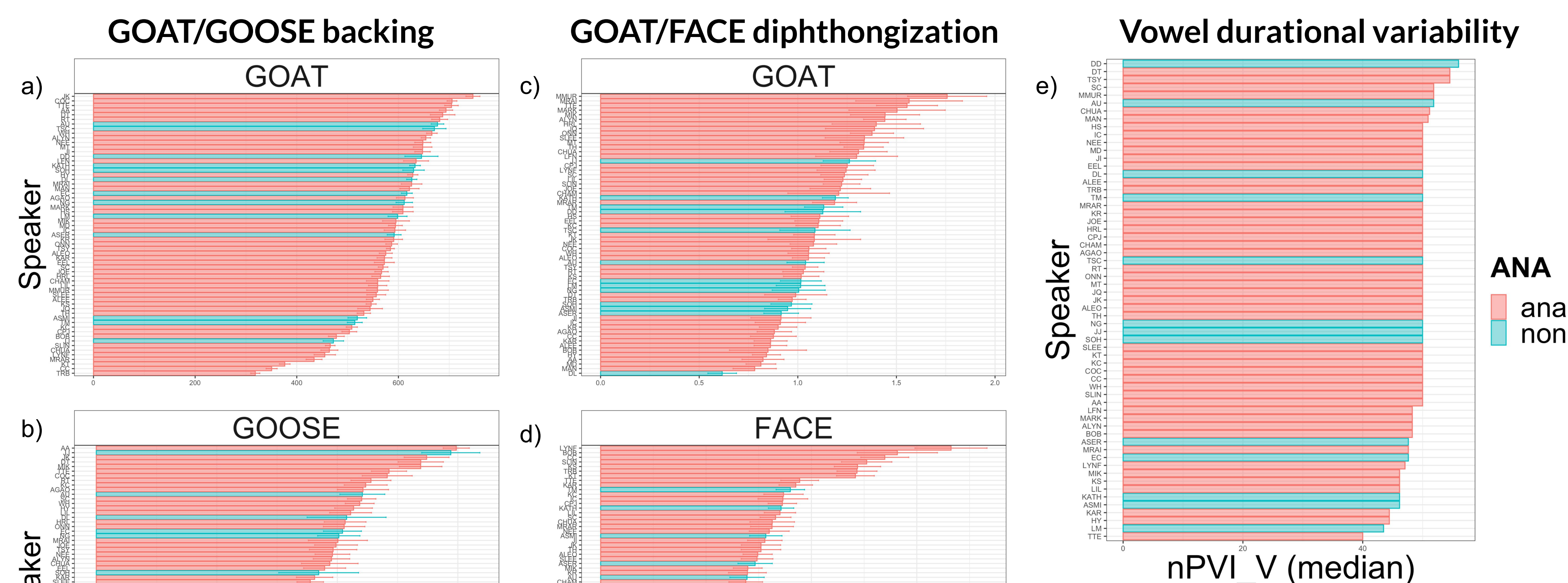


Fig 1. Values per measure plotted by speaker, sorted from highest to lowest. Color indicates ANA (red) and non-ANA (blue) identity.

Measures: Midpoint F2 difference from /i/ for (a) GOAT and (b) GOOSE. Norm. F1-F2 Euclidean distance between 25% and 75% for (c) GOAT and (d) FACE. (e) Normalized Pairwise Variability Index (Grabe & Low, 2002)

Limitations & Future Steps

- LingTube only thoroughly tested on MacOS and English speech, but easily extendible
- Still requires somewhat time-intensive semi-manual work, but planned upgrades will reduce this
 - i. Music/noise detection and removal
 - ii. Word-level auto-alignment (prior to phoneme-level forced-alignment)
- YouTube data may not be suitable for some questions
 - E.g. Hard to get ethnic and/or regional identity info
- Both auto-captioning and forced alignment may struggle with non-standardized American English varieties introducing bias into data
 - Current automatic vowel data needs correction
- Next steps:
 - Continued analysis with hand-corrected vowels, additional measures, consideration of more speaker information (e.g. region, age, etc.)
 - Possible extension to morphosyntactic analysis

References Bauman, C. (2016). Speaking of Sisterhood: A Sociolinguistic Study of an Asian American Sorority. Coats, S. (2020). Articulation Rate in American English in a Corpus of YouTube Videos. Grabe, E., & Low, E. L. (2002). Durational variability in speech and the Rhythm Class Hypothesis. Kramer, A. (2021). Dependency Lengths in Speech and Writing: A Cross-Linguistic Comparison via YouDePP, a Pipeline for Scraping and Parsing YouTube Captions. Lee, S. (2017). Style-Shifting in Vlogging: An Acoustic Analysis of “YouTube Voice.” McAuliffe, M., Socolof, M., Mihuc, S., Wagner, M., & Sonderegger, M. (2017). Montreal Forced Aligner: Trainable Text-Speech Alignment Using Kaldi. Schneider, E. W. (2016). World Englishes on YouTube: Treasure trove or nightmare?