



# 4aSCb8. Dynamic trajectories of tense vs. lax vowels in the American South

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## 1. Front vowel shifting in Southern American English

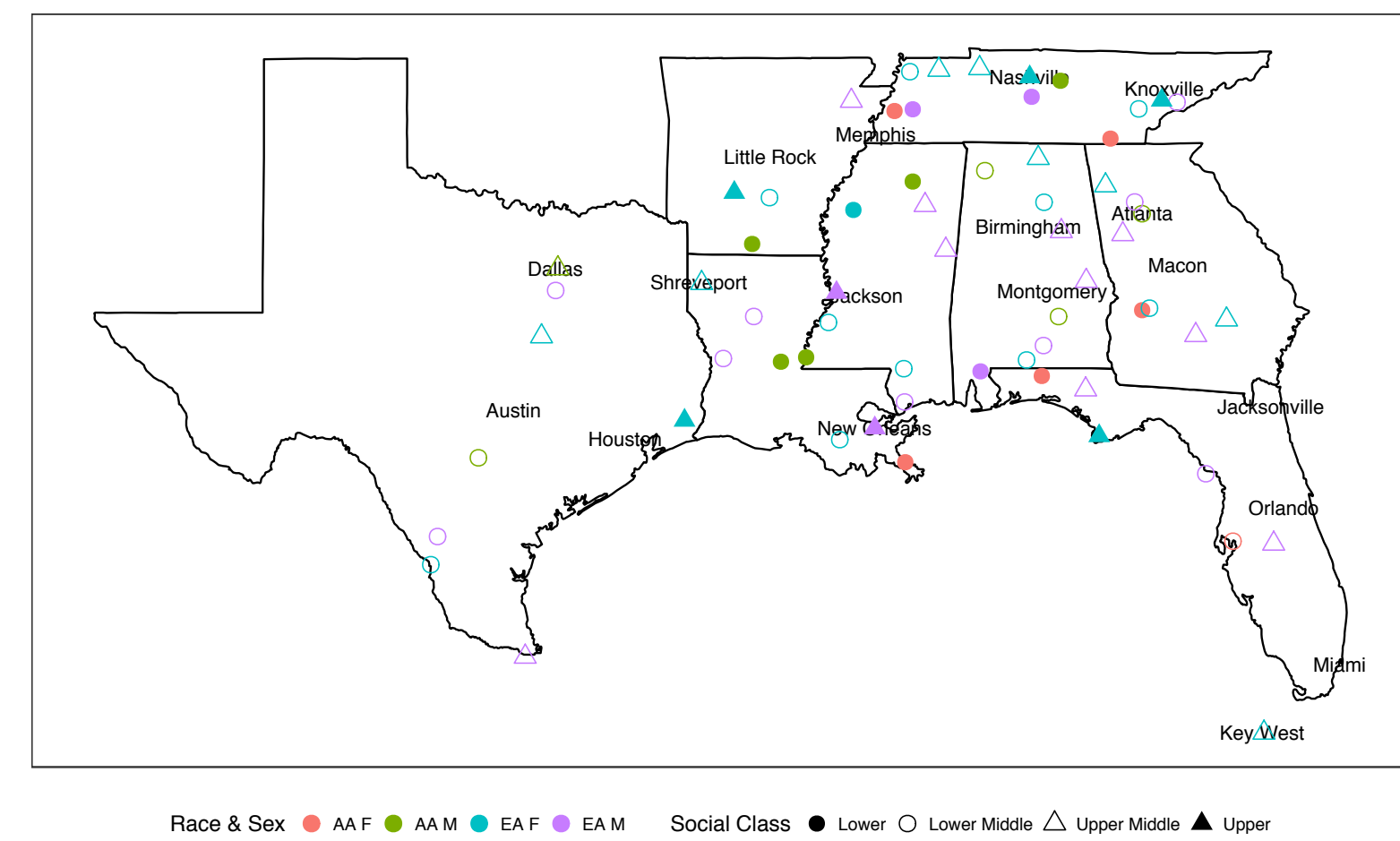
- Static measurements suggest that Southern front tense and lax vowels “swap places,” as in *beet-bit*, *bait-bet*. Given their strong acoustic overlap, what maintains their distinctness?
- Dynamic trajectories are shown to diverge in position and shape across tense and lax pairs.

### Research questions:

- How distinct are the trajectories of tense vs. lax vowels?
- Which socioindexical factors affect vowel trajectory shape?
- Can vowel trajectory modeling give insight into the spread of the Southern Shift?

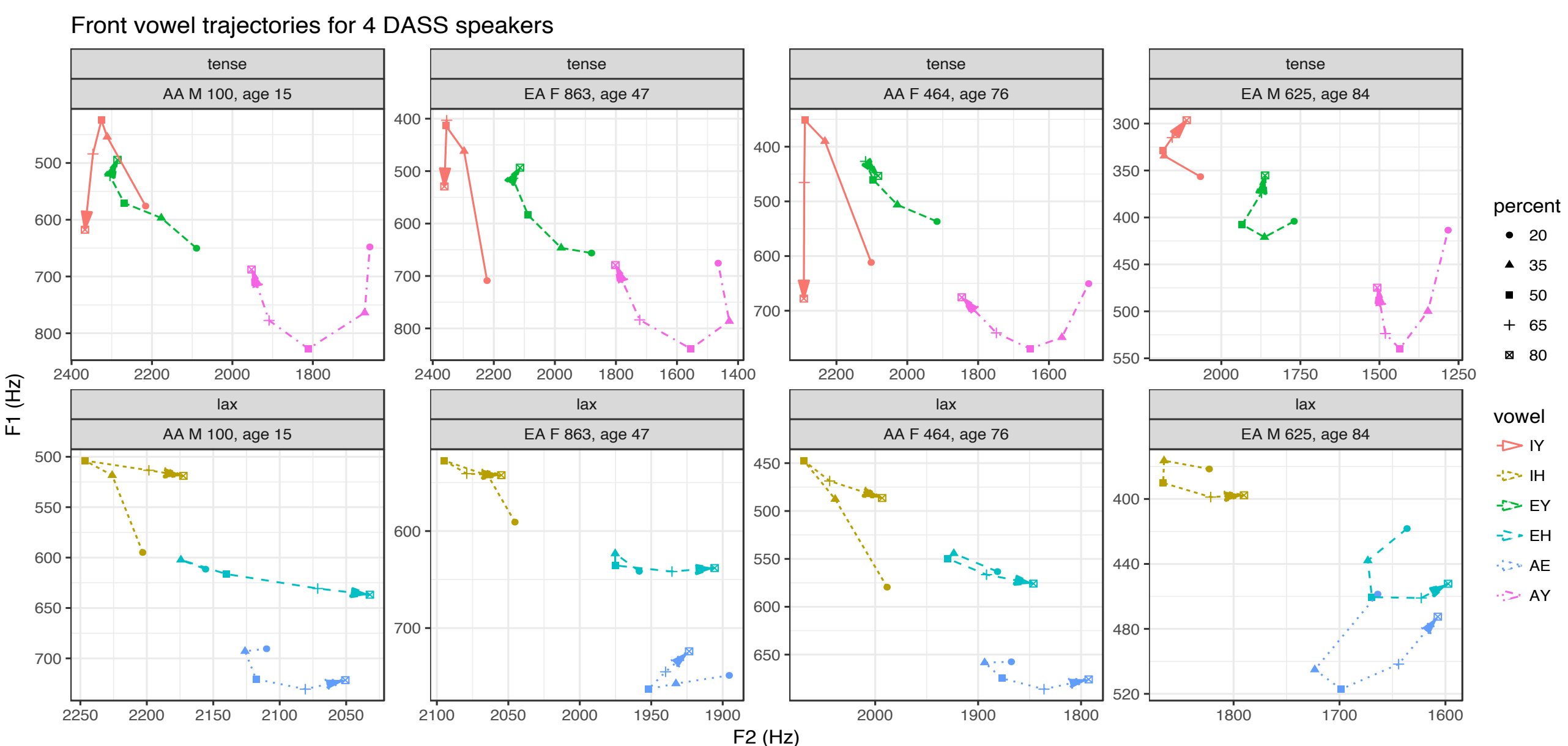
## 2. The Digital Archive of Southern Speech

- Audio corpus of semi-spontaneous linguistic atlas interviews (Kretschmar et al. 2012)
- 64 American speakers native to 8 Gulf States
- Speakers represent a mixture of ethnicities, social classes, education levels, ages
- DASS is being transcribed, aligned, and acoustically analyzed (Olsen et al. 2017)
- Data from **53 speakers** is currently available: 26 female (5 African American), 27 male (6 AA)

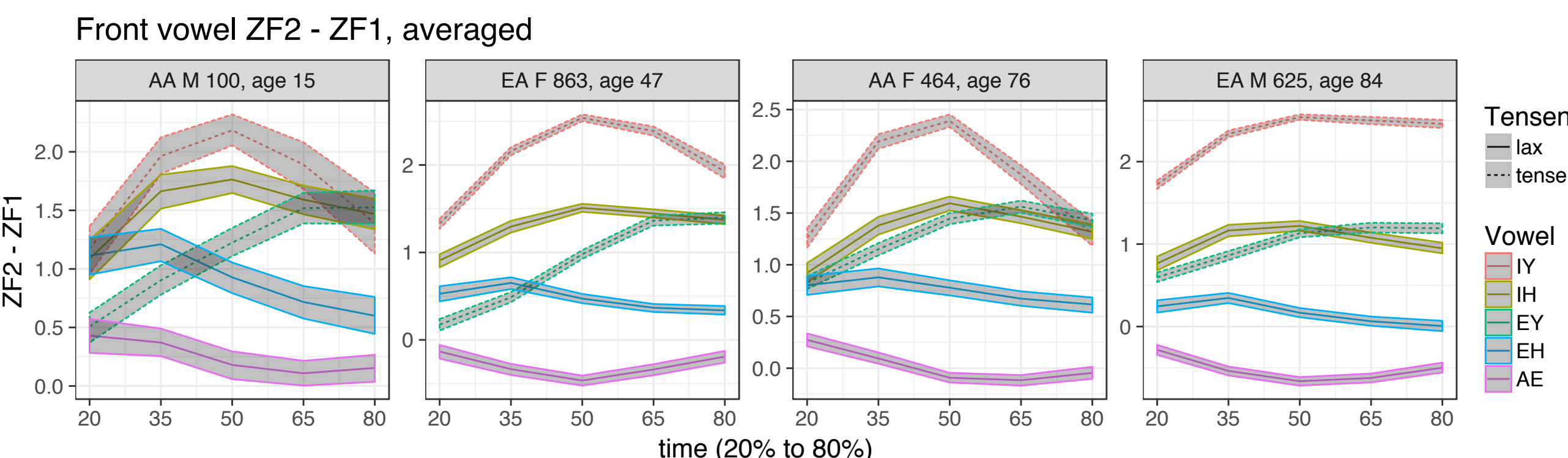


## 3. Data and data preparation

- Vowels:** 426,682 total tokens aligned and measured with DARLA (Reddy & Stanford 2015)
- Formant values: **F1, F2** extracted by DARLA at five time points: 20%, 35%, 50%, 65%, 80%
- Formant values normalized to speaker-specific z-scores (Lobanov 1971)
- Acoustic **filtering:** at the 50% F1, F2 values of each token, Mahalanobis distance (Mahalanobis 1936) is calculated, relative to a speaker- and vowel-specific centroid. Tokens with high distance (based on 95% quantile of a  $\chi^2$  distribution) are excluded as outliers
- Tokens **analyzed:** Stressed front vowels that do *not* precede nasals or liquids
- Position on vowel diagonal:** ZF2 – ZF1 is calculated for each vowel measurement (Dodsworth 2013) and plotted across time. Higher values indicate higher, fronter nuclei (Labov et al. 2013).



Vowel Symbol Tokens		
/i/	IY	17354
/eɪ/	EY	22863
/ɪ/	IH	15071
/ɛ/	EH	16896
/æ/	AE	23977
Total		96161



## 4. Generalized Additive Mixed Models

- GAMMs are useful for **dynamically variable** data (Soskuthy ms.)
- Linear predictors depend linearly on smooth functions (Wood 2017a); models can account for temporal autocorrelation (Baayen et al. 2017), and detect significant differences in smoothed curve shapes

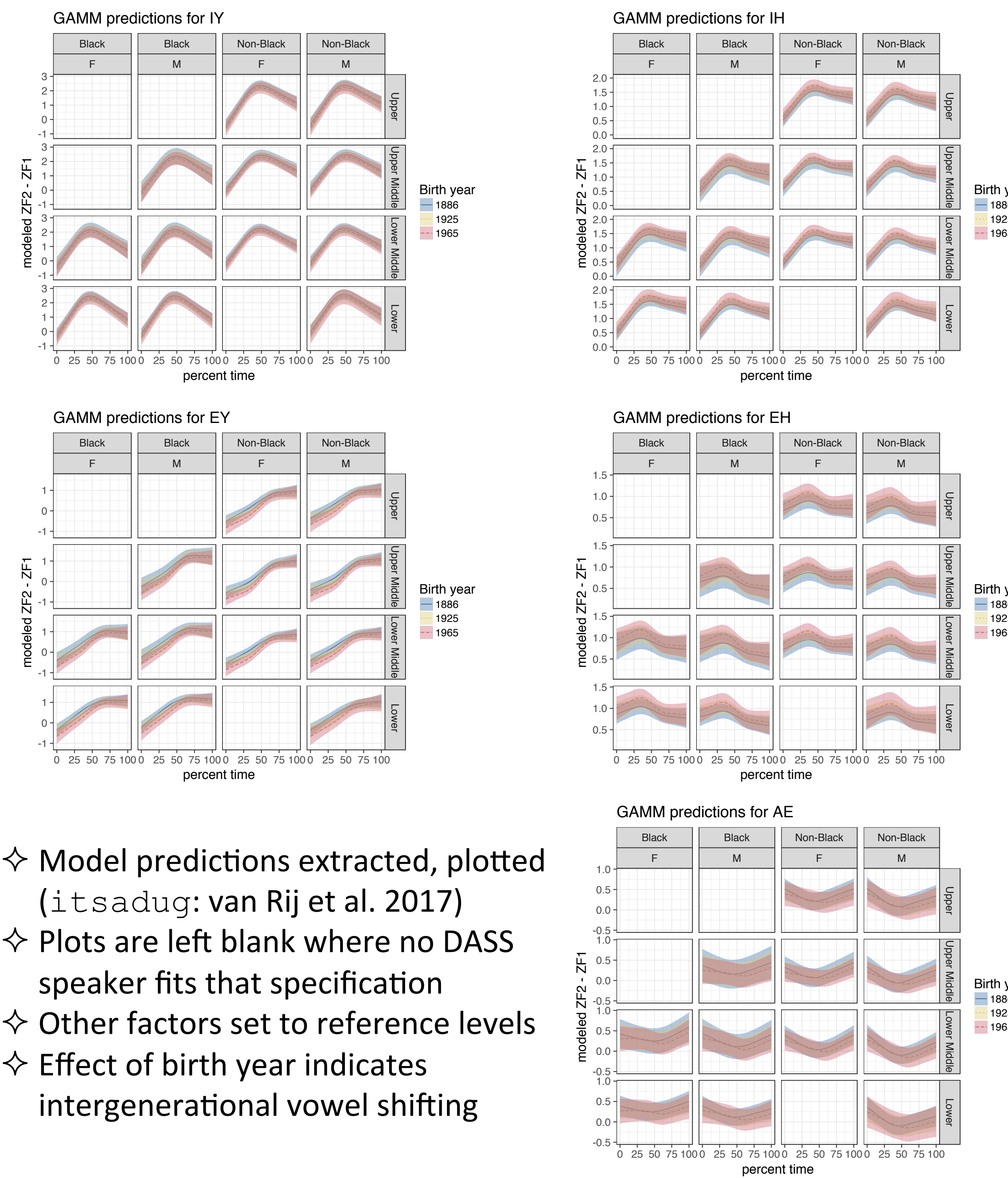
### GAMMs fit to vowel data

- Separate model fitted for each vowel (mgcv; Wood 2017b)
- Dependent variable: ZF2 – ZF1 (5 values per vowel token)
- Autoregressive error model seeded with  $\rho$  from identical ML model

### Predictor variables

- Phonological and random factors (not shown in Results table)
  - Following consonant place of articulation
  - Vowel context within word, indicating syllable structure
  - Vowel duration (smooth, and tensor product with Percent)
  - Speaker (random smooth)
- Socioindexical factors
  - Birth year (relativized: 1886 = 0; 1965 = 79)
  - Ethnicity (reference = Non-Black)
  - Sex (reference = Female)
  - Social Class: Upper (reference), Upper Middle, Lower Middle, Lower
  - State: AL (reference), AR, FL, GA, LA, MS, TN, TX

## 5. Estimated trajectories from fitted GAMMs



- Model predictions extracted, plotted (itsadug: van Rij et al. 2017)
- Plots are left blank where no DASS speaker fits that specification
- Other factors set to reference levels
- Effect of birth year indicates intergenerational vowel shifting

## 6. Results of statistical modeling

- Significance tested via conservative model comparison with dropped factors (compareML; van Rij et al. 2017)
- \* indicates factor is significant in model summary alone

Factor	IY	EY	IH	EH	AE
Birth year (relative)	*	$p < 0.001$	*	*	$p < 0.001$
Ethnicity	*	$p < 0.001$			$p < 0.05$
Sex		$p < 0.05$	$p < 0.05$	*	$p < 0.05$
Class	*	$p < 0.001$			
State	$p < 0.05$	$p < 0.01$	$p < 0.001$	$p < 0.01$	$p < 0.05$
Adjusted R <sup>2</sup>	0.278	0.29	0.173	0.132	0.205

## 7. Conclusions

- Tense and lax front vowels have distinct dynamic trajectories
- IY vs. IH**
  - IY starts higher, fronter than IH
  - IY has more extreme trajectory, finishes more centrally
  - Both vowels reach peak ZF2 – ZF1 near their 50% point
  - GAMM trends suggest IY lowering, IH raising across generations
- EY vs. EH**
  - EY is heavily centralized in first half, thus lower, backer than EH
  - EH is most central at its start and end
  - EY raises and fronts over its timecourse; shows greatest variation across time, ethnicities, sexes, classes and states
- AE**
  - AE is highest and most fronted in its first half
  - Over generational time, the shape of AE’s trajectory changes, and its lowest point shifts later in vowel
  - Young, African American females have a distinct trajectory suggesting increased raising, and strongest shifting for AE
- Future research will test precisely at what points, and factor levels, trajectories are significantly different

## Acknowledgments & References

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