### Phonology has an Early Influence on Sound Change

Josef Fruehwald, NELS 43 Model Description

### 1 Data

y is F1.

$$y_{1,2...n} \tag{1}$$

J is a vector of speaker indices.

$$J_{1,2...n} \tag{2}$$

$$J_{1,2...n} (2)$$

$$j = J_i (3)$$

K is a vector of context indices.

$$k = 1$$
 Surface /d/  
 $k = 2$  Surface /t/  
 $k = 3$  Flapped /d/  
 $k = 4$  Flapped /t/

$$K_{1,2...n} \tag{4}$$

$$K_{1,2...n} (4)$$

$$k = K_i (5)$$

W is a vector of word indices.

$$W_{1,2...n} \tag{6}$$

$$W_{1,2...n} \tag{6}$$

$$m = W_i \tag{7}$$

D is a vector of durations. Original msec measures have been log2 transformed and centered around the median.

$$D_{1,2...n} \tag{8}$$

$$d = D_i \tag{9}$$

(10)

B is a vector of dates of birth for each speaker.

$$B_{1,2,\dots,max(J)} \tag{11}$$

$$B_{1,2...max(J)}$$

$$b = B_j$$

$$(11)$$

$$(12)$$

(13)

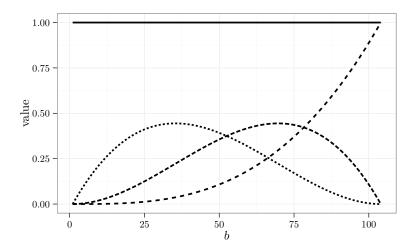


Figure 1: b-spline basis

x is the b-spline basis represented in Figure 1.

$$x_{1...max(B),1...4}$$
 (14)

### The Model $\mathbf{2}$

The change over time is modeled with the b-spline basis by multiplying it by a matrix of weighting coefficients,  $\beta$ .

$$\beta_{1\dots 4,1\dots max(K)} \tag{15}$$

$$\beta_{1...4,1...max(K)}$$

$$\gamma_{bk} = x \times \beta$$
(15)
(16)

We want to fit the following model represented in Figure 2 for every speaker.

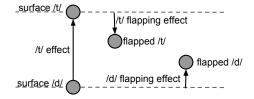


Figure 2: Speaker Model

$$\gamma_{jk}^s \sim \mathcal{N}(\gamma_{bk}, \sigma_k)$$
 (17)

$$\mu_{jk=1}^{s} = \gamma_{jk=1}^{s}$$

$$\mu_{jk=2}^{s} = \gamma_{jk=1}^{s} + \gamma_{jk=2}^{s}$$

$$\mu_{jk=3}^{s} = \gamma_{jk=1}^{s} + \gamma_{jk=3}^{s}$$

$$(18)$$

$$(19)$$

$$(20)$$

$$\mu_{jk=2}^{s} = \gamma_{jk=1}^{s} + \gamma_{jk=2}^{s} \tag{19}$$

$$\mu_{ik=3}^{s} = \gamma_{ik=1}^{s} + \gamma_{ik=3}^{s} \tag{20}$$

$$\mu_{jk=4}^{s} = \gamma_{jk=1}^{s} + \gamma_{jk=2}^{s} + \gamma_{jk=4}^{s}$$
(21)

We also want to estimate word-level effects.

$$\mu_m^w \sim \mathcal{N}(0, \sigma^w)$$
 (22)

We'll also estimate a duration effect,  $\beta^d$ , and speaker-level duration effects,  $\beta^{ds}_j$ .

$$\beta_j^{ds} \sim \mathcal{N}(\beta^d, \sigma^d)$$
 (23)

Finally, the data is estimated as,

$$y_i \sim \mathcal{N}(\mu_{jk}^s + \mu_m^w + (\beta_j^{ds} \times d), \sigma_j^s)$$
 (24)

Where,  $\sigma_j^s$  is a speaker specific dispersion parameter.

Any parameters for which a prior has not been explicitly defined in this description was either given a  $\sim U(0, 100)$  in the case of variance parameters, or  $\sim \mathcal{N}(0, 1000)$  for all others.

### 3 **Implementation**

This model was estimated by Hamiltonian Monte Carlo, using Stan.

# Phonology Has an Early Influence

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# on Sound Change



University of Pennsylvania

### Which comes first?

phonological change the accumulation of gradient Does gradient phonetic change feed subsequent categorical phonological change? Is apparent phonetic errors in production or perception?

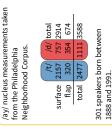
## Test Case: /ay/ Raising

Normalized F1

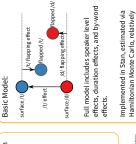
/ay/ raises in Philadelphia before voiceless consonants only, and exhibits opacity in contemporary speech.



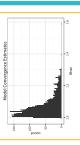
### Data and Model

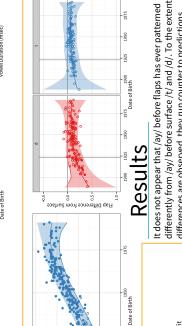


### References



well converged.





differences are observed, they run counter to predictions At all times in the change, /ay/ raising has occured to a degree proportionate to the underlying voicing of the following segment, not proportionate to the phonetic properties of its context. based on phonetic bias.

A model where /ay/ raising began due to phonetic biases, then generalized along phonological lines is not supported. Rather, the phonological generalization appears to be concurrent with the phonetic shift.