

LSA 511: Computational Models of Sound Change

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Why are we here?

Frequency effects and their relation to lenition (weakening)

- Frequency often correlated with **synchronic** reduction (Hooper, 1976; Bybee, 2000; Jurafsky et al., 2002; Munson and Solomon, 2004...)
 - *memory* > [mɛmɹi], but *mammary* ✗ [mæmɹi] (Hooper, 1976)
 - *told* > [tɒl] but never *meant* > [men] (Bybee, 2000)
- Frequency also correlated with rates of **diachronic** change (e.g. Philips, 1984)
 - (low-frequency) *nude* */njud/ > [nu:d], not [nju:d], while
 - (high-frequency) *new* */nju/ > [nju:], not [nu:]

Why are we here?

- Goal: an **existence proof** by constructing 'a formal architecture which is capable of capturing these regularities'
- Should also predict that 'some outcomes are possible and others are not'

Pierrehumbert (2001)

‘By examining the consequences of the perception-production loop over time, we provide a formal framework for thinking about the quantitative predictions of usage-based phonology, as proposed by Bybee. **We derive the finding that leniting historical changes are more advanced in frequent words than in rarer ones.** Calculations are presented which reveal the interaction of production noise, lenition and entrenchment. **A realistic treatment is also provided for the time course of a phonological merger which originates from lenition of a marked category.**’

Pierrehumbert (2001)

Other (tacit/explicit) claims made by/attribution to this paper:

- Under certain conditions, **category variance will stabilize**
- **Frequency** plays a role in neutralization

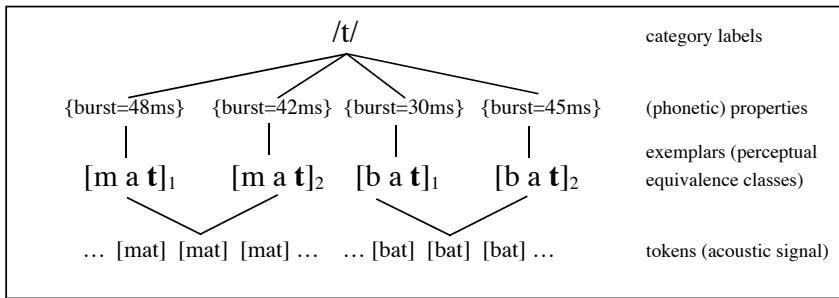
Things to keep in mind

1. Knowledge state of individuals
2. Network/social structure
3. Assumptions about communication
4. Learning algorithm

Exemplar 'theory'

- Like 'sound change', a potentially ambiguous term
- Memory is episodic, highly detailed
- System is a map from points in multidimensional phonetic space to category labels...
- ...independent of where those labels might come from

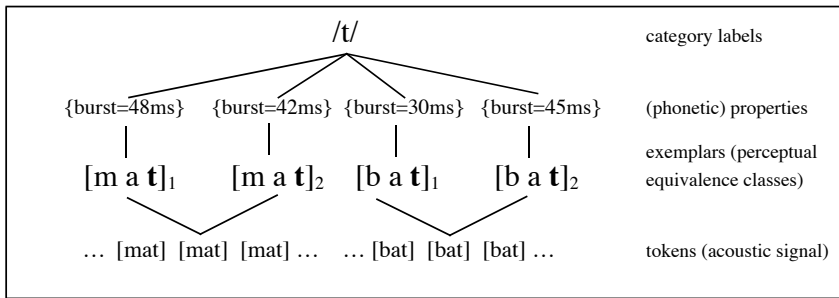
Exemplar storage: Units (knowledge state)



- An **exemplar list** $E(L)$ is a list of exemplars associated with label L :

$$E(L) = e_1^L, \dots, e_n^L$$

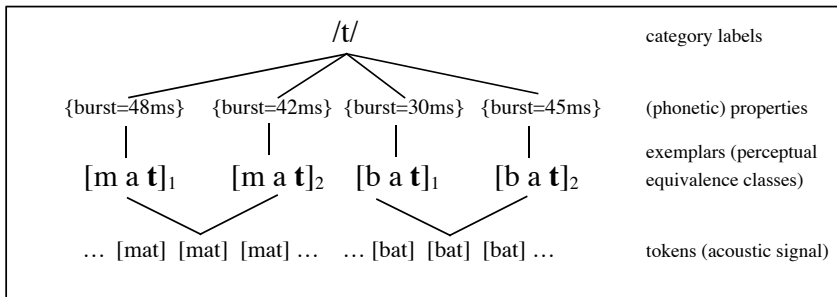
Exemplar storage: Units (knowledge state)



- Individual exemplars represented as (property, label) pairs

$$e_1^L = \{burst, 48ms\}, e_2^L = \{burst, 42ms\}, \dots$$

Exemplar storage: Constraints

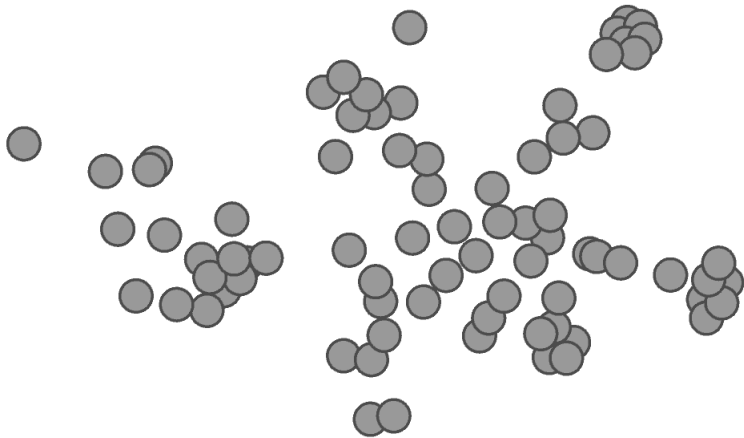


- Memory limitations \rightarrow memories **decay**
- Perceptual limitations \rightarrow **granular** representations

Model 1: Single category

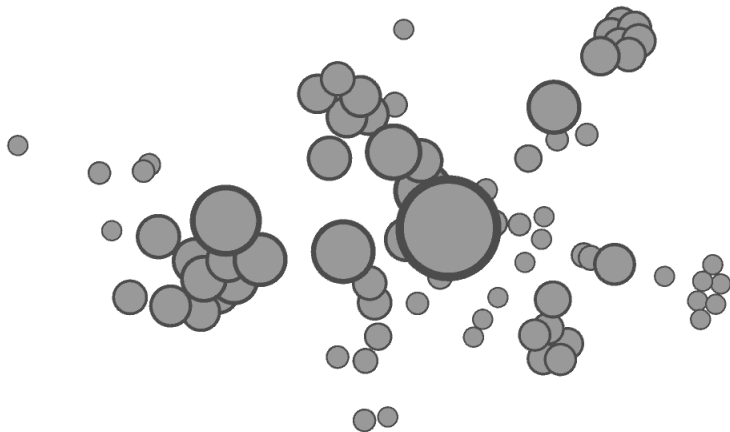
- Single speaker/hearer, talking to themselves
- Entails simple(st) network structure
- Knowledge state as detailed above
- What are the assumptions about communication and learning?

Model 1: Production



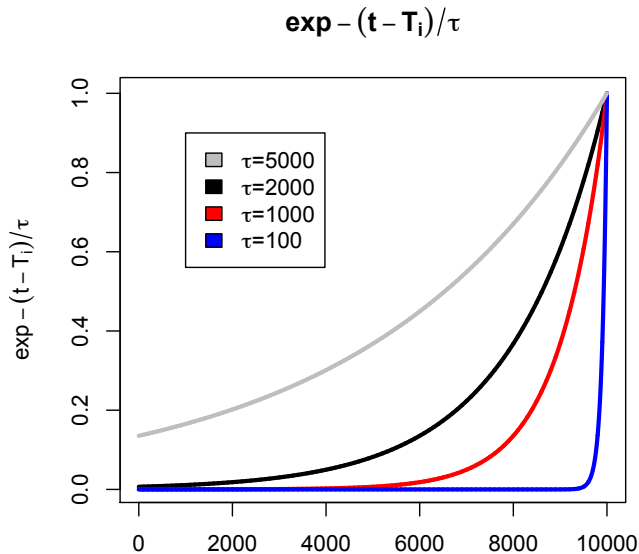
...if target were selected with **uniform** probability

Model 1: Production



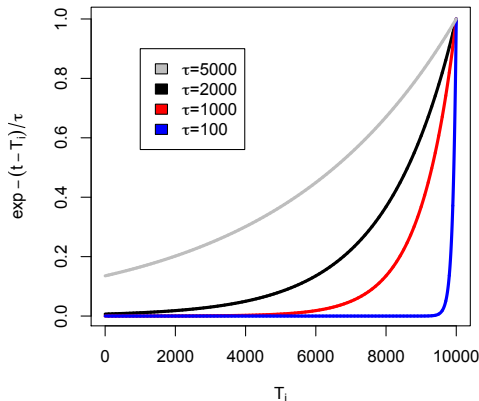
...target selected **weighted by recency**: $\exp(-\frac{t-T_i}{\tau})$

Model 1: Production



Model 1: Production

$$\exp - (t - T_i)/\tau$$



- Resting activation level a function of time and τ :

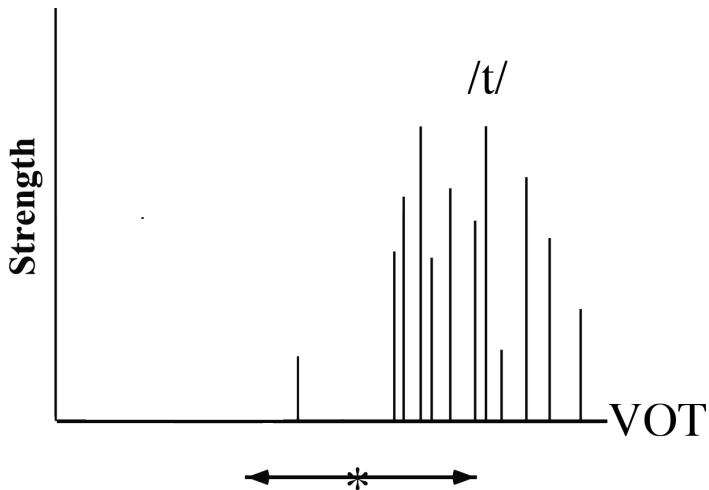
$$\exp\left(-\frac{t - T_i}{\tau}\right)$$

- Bounded between 0 and 1
- As numerator gets **bigger**, weight gets **smaller**

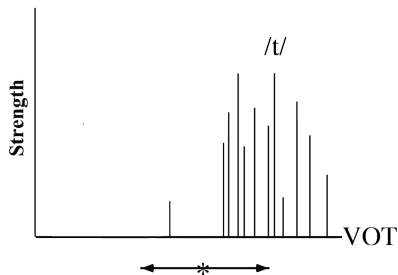
Model 1: Production

- **Noise**: ‘random deviations from the phonetic target due to noise in the motor control and execution’
- Modify production target $x = x_{target} + \epsilon$
- ϵ : random number chosen from a **uniform** distribution over some fixed range (e.g., from -0.1 to 0.1)

Model 1: Perception



Model 1: Perception

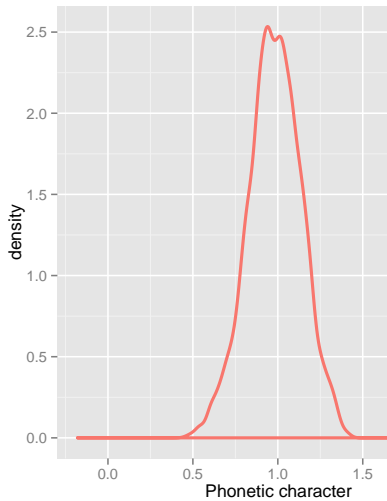


$$W(x - e_i^L) = \begin{cases} 1 & \text{if } |x - e_i^L| < 0.05 \\ 0 & \text{otherwise} \end{cases}$$

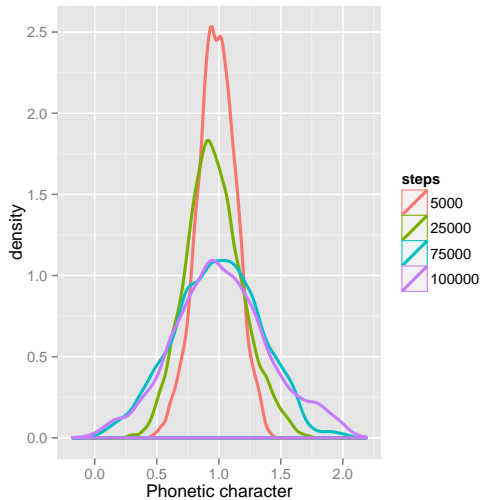
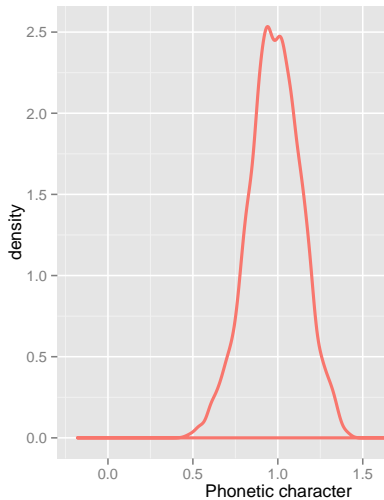
$$\text{score}(L, x) = \sum_{i=1}^n W(x - e_i^L)$$

If $\text{score} \neq 0$, x is retained. (learning!)

Model 1: Results



Model 1: Results

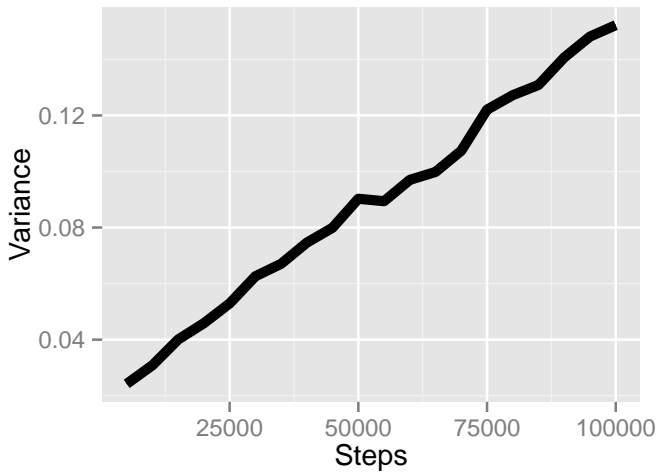


Model 1: Results

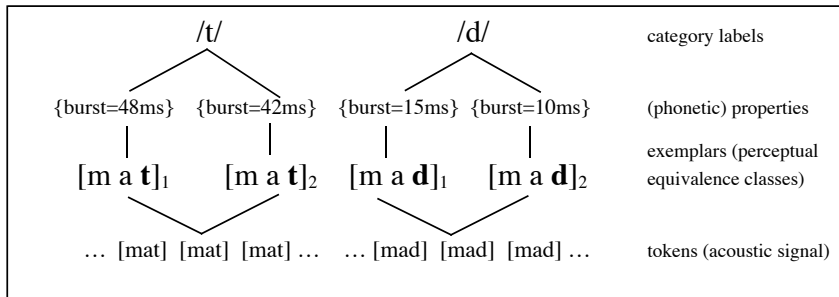
What does P conclude from this?

- 'The total representation of the category is strengthened as more and more memories are stored'
- 'the variance of the distribution ... increases with usage'

Model 1: Results



Model 1a: Two categories

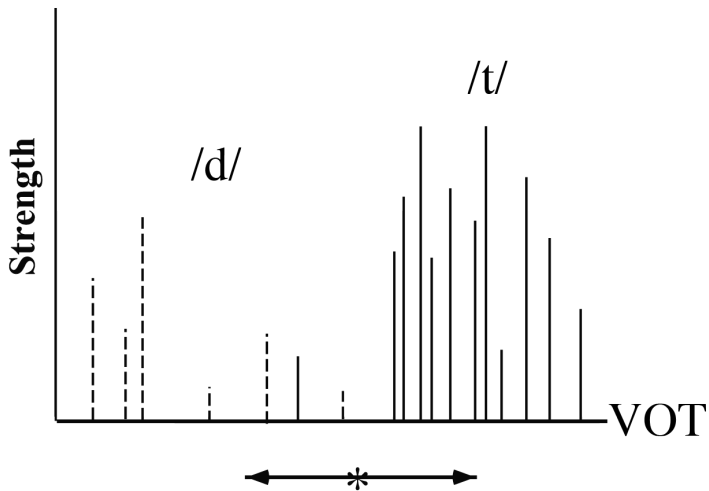


- How do we extend this model to more than one category?

Model 1a: Production

- Now that there are two categories A and B , which one does the speaker utter an instance of at any given timestep?
- Each list seeded w/ a single value; then randomly produce a token x of A or B with probability p and $1 - p$, respectively
- p is fixed/static: e.g. phrase-final voiced obstruents (B) are *a priori* less likely than phrase-final voiceless obstruents (A)

Model 1a: Perception



Model 1a: Perception

- New exemplars are given a **score** for each list (category) L :

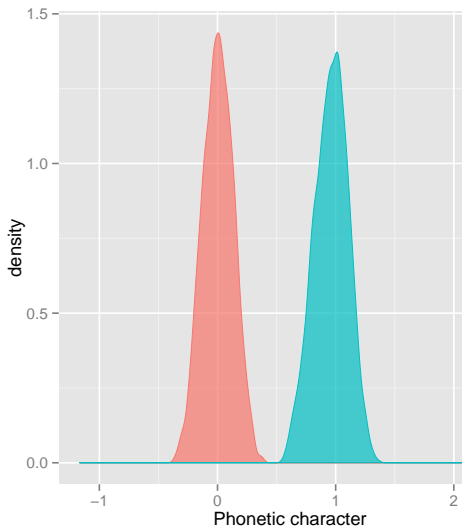
$$score(L, x) = \sum_i^n \underbrace{W(x - e_i^L)}_{\text{distance}} \overbrace{\exp(-\frac{t - T_i}{\tau})}^{\text{activation level}} \quad (1)$$

- Assign input x to list with the highest score:

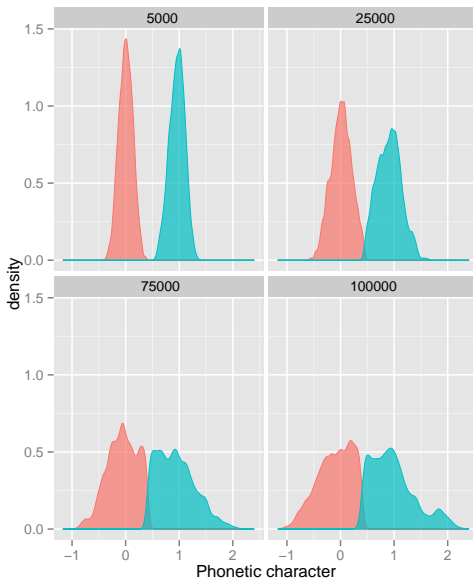
$$\arg \max_L score(L_n, x), L \in \{L_1, \dots, L_n\} \quad (2)$$

- If $L_i = L_j$ for all i, j , **discard** x .

Model 1a: Results



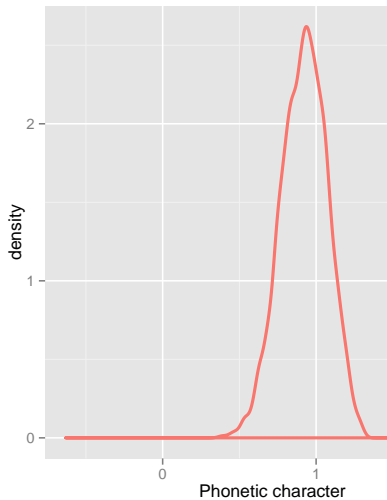
Model 1a: Results



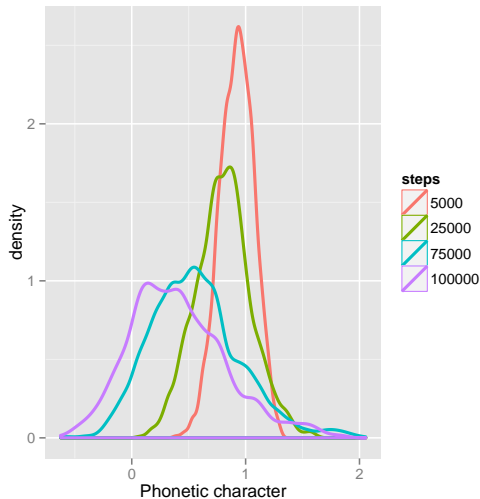
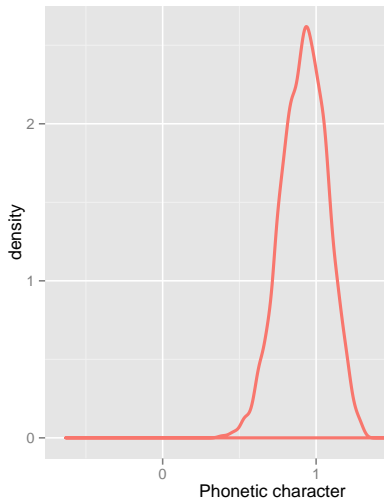
Model 2: Bias

- Motivation: ‘tendency to undershoot ... in order to save effort and speed up communication’ (Lindblom, 1984)
- Drives leniting changes like schwa reduction and /t/-deletion?
- Production target is now $x + \epsilon + \lambda$
- λ is a **constant** (i.e. fixed) parameter (e.g., -0.1)

Model 2: Results



Model 2: Results



Model 2: Discussion

- 'We derive the finding that leniting historical changes are **more advanced in frequent words** than in rarer ones....'
- In what sense does this model derive this?
- Synchronic interpretation?

Model 3: Lenition + Entrenchment

- **Problem:** variance appears to increase without bound
- Aggravated by lenition bias, but...
- Phonetic variability *decreases* with practice (Lee et al 1999)
- 'There is **no combination of parameter settings** for the model which allows a category to fill out after being seeded by a single example, without simultaneously predicting that the spreading out will go on indefinitely'

Model 3: Lenition + Entrenchment

- **Solution**: entrenchment
- **Production** involves **averaging over multiple exemplars**
- Pick n_{trench} closest exemplars to target, where ‘close’ is the **inverse** of the weighted distance used in perceptual scoring:

$$d_i = |x_{target} - e_i^L| \exp\left(\frac{t - T_i}{\tau}\right) \quad (3)$$

$$\omega(T_i, t) = \exp\left(\frac{t - T_i}{\tau}\right) \quad (4)$$

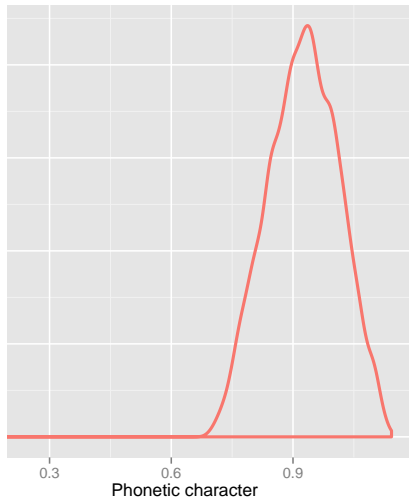
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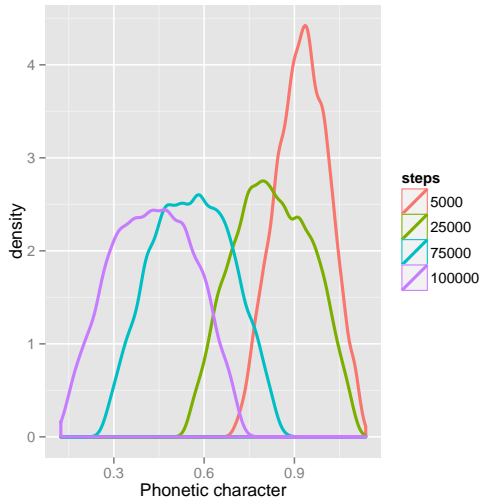
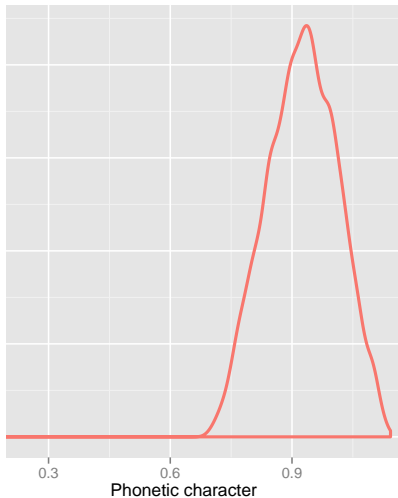
$$d_i = |x_{target} - e_i^L| \omega(T_i, t) \quad (5)$$

$$x = \frac{\sum x_i \omega(T_i, t)}{\sum \omega(T_i, t)} \text{ for } n_{trench} \text{ closest exemplars} \quad (6)$$

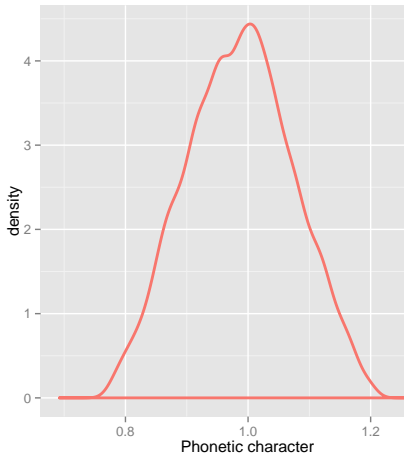
Model 3: Results



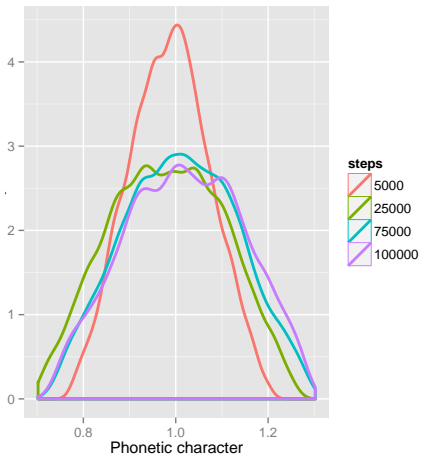
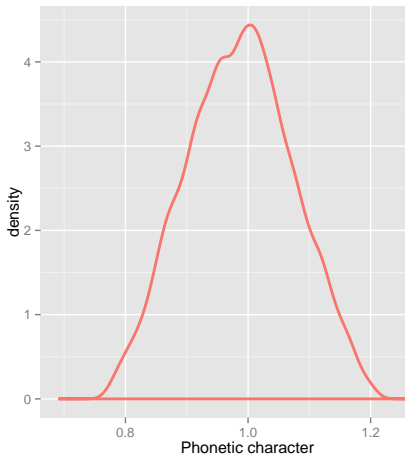
Model 3: Results



Model 3a: Entrenchment only



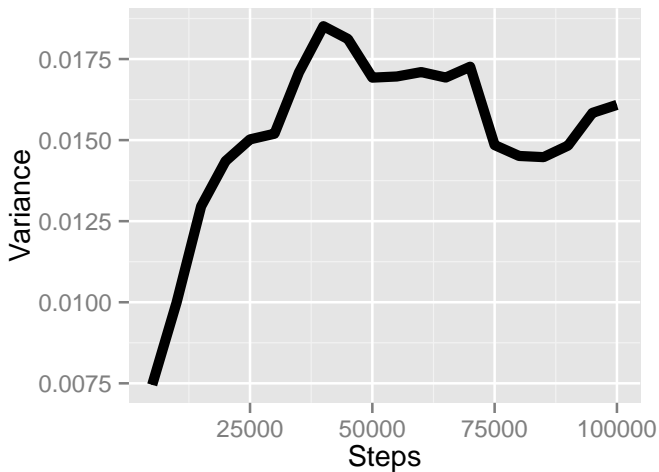
Model 3a: Entrenchment only



Models 3/3a: Conclusions

- 'With the particular parameter settings selected here, the spreading effects arising from production noise and lenition and the anti-diffusive effect of entrenchment have **essentially cancelled out** in determining the variance.'
- In other words: variance **stabilizes** at some point

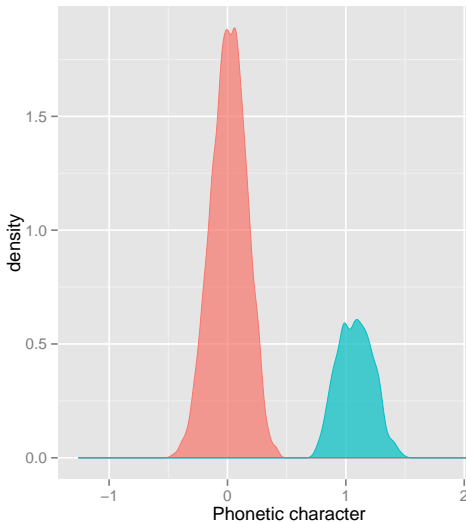
Models 3/3a: Entrenchment



Model 4: Neutralization

- **Intuition:** lenition bias in a region of phonetic space shared by multiple categories will eventually result in **stable merger**
- ‘We consider the case of a marked phonological category competing with an unmarked one....we take the unmarked category to be **more frequent** than the marked one...’
- P shows us lenition *plus* entrenchment *plus* unequal frequency
- What about just unequal frequencies, full stop?

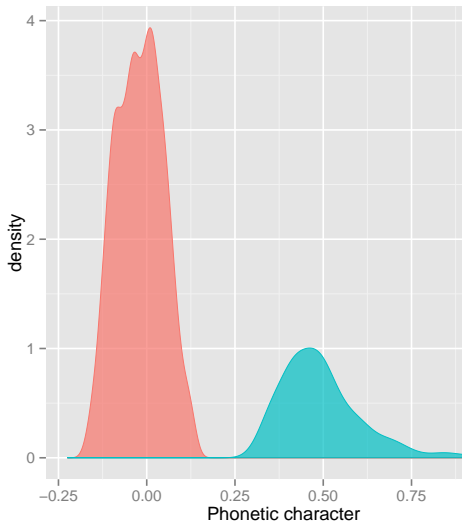
No lenition, no entrenchment, unequal frequencies



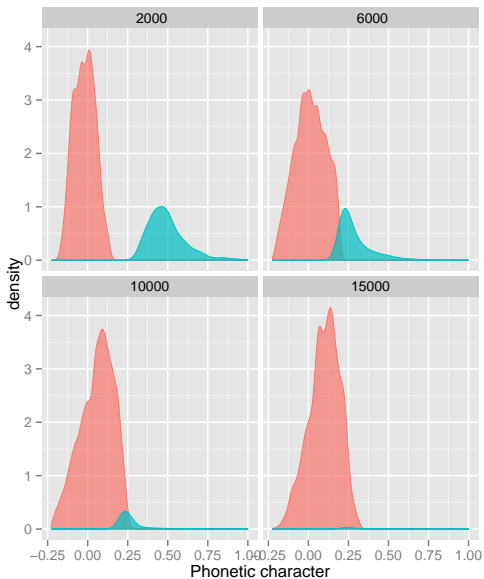
No lenition, no entrenchment, unequal frequencies



Model 4: Results



Model 4: Results



Model 4: Conclusions

- Is this a 'realistic treatment of the time course of phonological merger'?
- If it is, what should the time course of change look like?

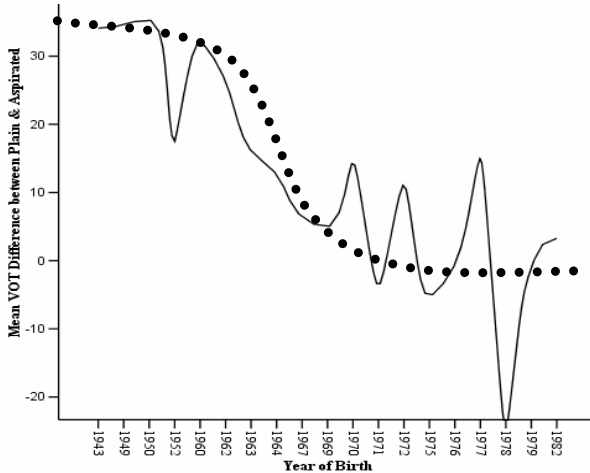
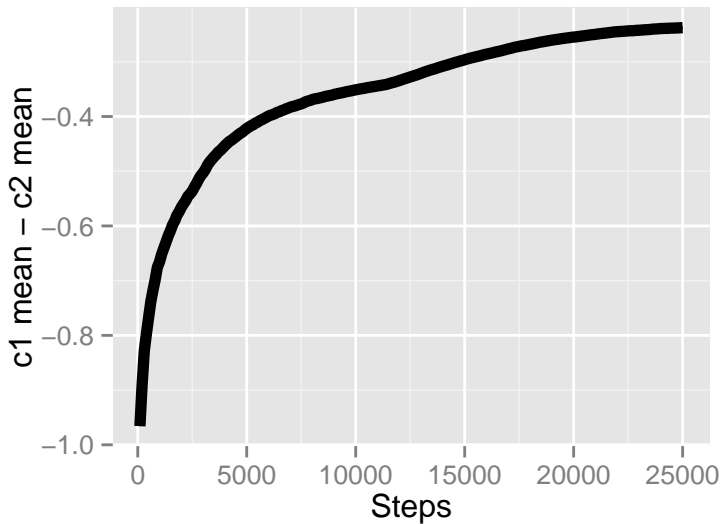


Figure 3. Age-related change for the VOT Gap (ms) ($\text{mean } VOT_{asp} - \text{mean } VOT_{lax}$).

Model 4



Model 4

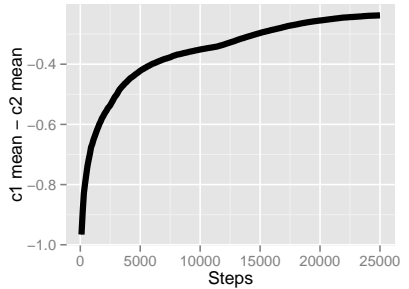
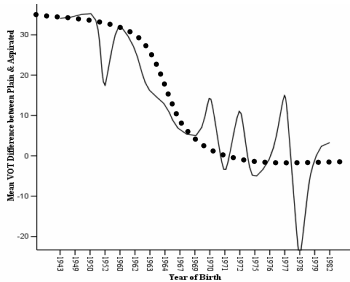


Figure 3. Age-related change for the VOT Gap (ms) ($\text{mean VOT}_{\text{asp}} - \text{mean VOT}_{\text{lex}}$).

Assumptions/limitations

- **Category labels.** Here: given. Where do they come from?
- **Unidimensionality of phonetic character.** What would be required to move to a multidimensional representation?
- **Network/social structure.** Realism of modelling single speaker-hearer?
- ...

What have we learned?

- (Non-)existence proof(s)?
- Explicitness?
- Counterintuitive results?
- Qualitative predictions?
- Baseline?

Weekend: extensions and explorations

- **Frequency and lenition.** What is necessary for neutralization to obtain: differences in frequency (p), lenition, or both?
- **Entrenchment: one category.** Does the variance always stabilize with nonzero entrenchment? Evaluate the claim that stable variance decreases as n_{trench} increases.
- **Entrenchment: two categories.** Vary the amount of entrenchment in the neutralization simulation. What happens?
- **Window size.** How does changing the window size impact the evolution of categories in the two category cases?

Weekend: extensions and explorations

- **Changing τ .** How does changing τ effect the evolution of the mean and variance of a category over time?
- **Noise.** Is there an effect of increasing/decreasing the amount of noise (i.e., changing endpoints of the distribution from which noise (ϵ) is drawn)?
- **Lenition.** What happens in the neutralization scenario as the amount of lenition λ is varied?
- **Advanced topics.** 3 categories? Alternative scoring function?

Post guidelines

- What did you try?
- What did you find? (pictures optional)
- Can you give an intuitive explanation for why you got the result that you did?
- What empirical/linguistic prediction(s) does your finding make?

Post guidelines

- 1 member of each group: please **email us group list** ASAP
- Example Piazza post, code available tomorrow
- Posts due **5pm Monday 1 Jul**