MODELLING SIMILARITY-DRIVEN PHONOLOGICAL GENERALISATIONS **Deepthi Gopal¹, Henri Kauhanen^{1 2} & Stephen Nichols¹** ¹University of Manchester & ²Universität Konstanz

We investigate:

- i. guesses at the trajectory and actuation of a change based on a *narrow* apparent-time sample;
- ii. the descriptive facts for a novel set of observations of the Turkish vowel space; mechanisms generalising a phonological rule to a class *and* producing an unequal but opposed effect in a different class;
- iii. the amount of information we'd have to give the most minimal possible 'toy' iterative learner in order for it to evolve in a qualitatively convincing manner.

Some real-world data

In Turkish, an apparently categorical phonological process lowers the front mid vowels /e/ and /ø/ to [x] and [œ] in pre-sonorant {r, l, m, n} contexts. (We focus here on the state of /e/, which is significantly higher-frequency overall and for which the data are more abundant.) Data (production study, n=10; all female speakers, age range 20–39, from major metropolitan areas in Turkey) show several interesting properties (for a detailed discussion, esp. of exceptionality, see Gopal & Nichols 2016 & in prep), & represent the first systematic investigation:

- i. Clear bimodality in F1/F2 space: realisations of /e/ in pre-sonorant and other contexts are nonoverlapping.
- ii. A further effect of coda sonority appears in pre-sonorant /e/: F1[r] > F1[l] > F1[n], F1[m]
- iii. Pre-obstruent /e/s are generally higher (smaller F1, larger F2) than /e/s in unchecked syllables. Preobstruent /ø/s are generally *lower* than /ø/s in unchecked syllables.



Figure 1: (Lobanov-normalised) F1/F2 space across all subjects, by vowel and by coda type.



Figure 2: F1/F2 space, showing 95% confidence ellipses. Left: underlying /e/ (1529 tokens), by coda type. *Right:* underlying /ø/ (332 tokens), by coda type.

When we search for birth-year effects (with the necessary caveats emptor on sampling and x-axis sparseness): i. pre-obstruent /e/-realisations begin to diverge noticeably from others with time (near-linearly); ii. the rate of change for /e/ is quite small for other contexts; the pre-sonorant /ø/s are moving faster;

- iii. for /e/, individual sonorant contexts' separation also generally time-invariant.
- iv. Overall: /ø/ is behind /e/. Open syllable raising in /ø/ disappears in /e/; pre-sonorant /ø/-lowering is less drastic, and distributions still show overlap; the pre-obstruent /ø/ trajectory begins to diverge from the pre-sonorant one.

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'you' [s**æ**n] [se.nin] 'your' [ar.darm]'virtue' [gæl.mek] 'to come' 'four' [dœrt 'see!' [gœr] 'file' [tœr.py [t∫øp] 'rubbish' syllable_type • obstruent open sonorant



Figure 3: Trajectories across 20 years' apparent time, using F2-2*F1 to model movement along the front diagonal of the vowel space. Dashed line tracks overall (cross-category) mean; generalised linear model fit. Top: left /e/, right /ø/. Bottom: Sonorants only, further split by individual coda; left /e/, right /ø/. NB: /ø/ tokens are <u>not</u> evenly distributed across coda environments – we avoid overinterpretation.

r is special: pre-/r/ tokens are consistently lower; for speakers who are far further behind in the change (see Gopal & Nichols 2016) nonetheless have pre-/r/ lowering. Claim: mid-vowel lowering is ultimately rtriggered, and generalises to the sonorants. Compare northeastern Swiss German varieties (Keel 1982; Janda & Joseph 2001, Mielke 2008), in which $o \rightarrow o$ generalises from pre-r to pre-sonorant and other. What's different here: the further distribution by individual coda segment, and the pre-obstruent raising.

Suggestion: One way to do this: generalisation is driven by a pressure to equalise across similar environments (the set of sonorants is fairly self-similar), and a pressure to avoid dissimilar environments (obstruents and sonorants disagree). No-coda neutral – absent pressure to raise in open syllables, drift towards the mean.

Some imaginary-world data

Let's consider a learner (a speaker-listener loop) whose capacities are quite limited: the only quantity it can observe and manipulate is a parameter corresponding to vowel height $b \in [0, 1]$. On each iteration, this learner: samples input distribution of /e/ heights \rightarrow adjusts this distribution with some biases \rightarrow produces output distribution Initialisation. Extrapolating from the production data: the 'initial state' involves some raising in unchecked syllables, and no other significant cross-context variation. Turkish-like context probabilities.

Biases. We provide our learner with:

i. a non-bias: default to the 'UR', or mean across all input tokens. ii. a 'phonetic' bias: at each iteration, a fraction of the set of pre-r tokens is adjusted downward;

iii. an operational-phonological bias: for each environment, the learner wants to converge towards other similar environments, and diverge from dissimilar ones.

We (optionally; see fig. 5) introduce a further bias parameter as follows: the listener may privilege comparison against the r context (is more 'aware' that pre-r tokens constitute an independent set). Does this affect its behaviour? Converges towards extreme states faster, but qualitative behaviour isn't very different. Suggestion: a more 'realistic' learner may vary such a parameter as it perceives that context separation has increased.

At each iteration *i*, the *next* height parameter $b_c[i+1]$ for context *c* is given by:

$$(1 - n)UR + n \sum_{q \in \text{ contexts}} p_q \cdot \sin(c, q) \cdot b_c(q)$$

where η determines conservativity of the learner ($\eta = 0$ blindly follows the across-context mean; $\eta = 1$ ignores it entirely), p_q is the probability that context q appears in the overall distribution of tokens, sim(c, q)'similarity score' $\in [-1, 1]$ for contexts *c* and *q*; $b_c(1 - b_c)$ forces the floor and ceiling [0, 1]; $(b_q - b_c)$ is the difference in context heights, which we aim to maximise if dissimilar and minimise if similar.

An elementary similarity-biasing function

 $(1-b_c)(b_q-b_c)$







Figure 5: Model averaged over 25 runs: parameter variation. Top row: Evolution over 10, 100, 200 iterations. Centre row: Evolution for values of 'phonologised' bias towards evaluation wrt. r. Bottom *row:* Evolution for values of conservativity/towards-mean parameter η .

We:

- Turkish, and of an opposite effect of raising for obstruents;
- apparent-time sample;
- that gives a qualitatively convincing trajectory vis-à-vis our data.

Remark. We see (in prep) inter-speaker variation on the status of coda voiced obstruents (low-frequency in Turkish – some generalise lowering to /v/, /z/ – model can account for this; see e. g. Mielke 2008, Cristia et al. 2013 on how this may support (featural?) similarity-, rather than class- generalisation.

Figure 4: 100 iterations of our model, averaged over 25 runs. Obstruent environments have been col-

i. have experimental evidence of a categorical lowering of /e/ (and, incipiently, /ø/) before sonorants in

ii. suggest aspects of the diachronic trajectory of this change that can be triangulated from a limited

iii. have presented a simple computational model of the speaker-listener loop based on assessed similarity