

Using Locality to Learn Long-distance Phonological Processes

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Main objectives

- Define *Output Strict Locality* (OSL), a restrictive and defining computational property of phonological $UR \mapsto SR$ maps.
- Analyze various long-distance processes as the composition of OSL maps.
- Demonstrate how the OSL property can be used to learn phonological processes.

Processes as Maps

- Phonological processes are *maps* from an input (UR) to an output (SR) (cf. Tesar 2014).
- We want to identify the best (i.e., most restrictive) characterization of the class of maps that is needed for phonology.
 - Why? Identifying this class helps us understand 1) what is a 'possible' phonological process and 2) how processes are learned.

Learnability

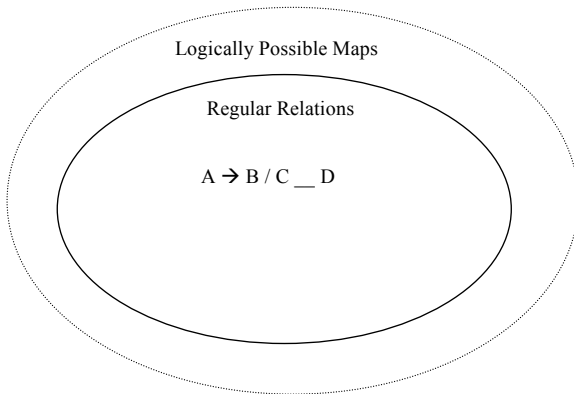
For a given class of maps, we want to know

- 1 Is there an algorithm that we can prove will learn any of the maps in the class?
- 2 Can it do so *efficiently* (i.e., with a reasonable amount of time and resources)?
- 3 What is the nature of the data it needs to learn the maps?

Methodology

- 1 Identify the defining property of the class of maps.
- 2 Prove that an algorithm that uses this property as an inductive principle will learn *all* and *only* the maps in that class.
- 3 Implication: no other maps can be learned and will therefore not be attested (i.e., the learner *is* the hypothesis space!).

Phonological maps are Regular Relations



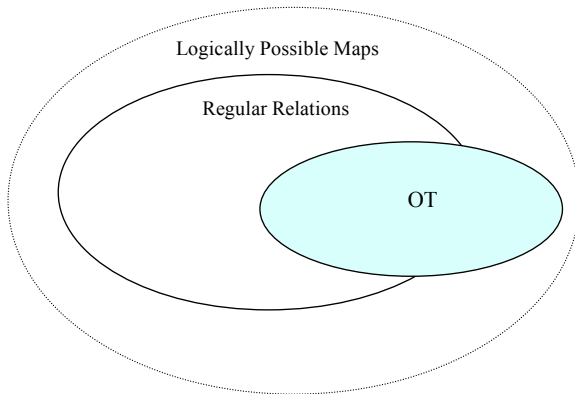
Johnson (1972); Kaplan and Kay (1994); Koskenniemi (1983)

Phonological maps are Regular Relations

- $A \rightarrow B / C \text{ — } D$
- $*CAD \gg \text{FAITH}(A \mapsto B)$

Baković (2013)

Phonological maps are Regular Relations

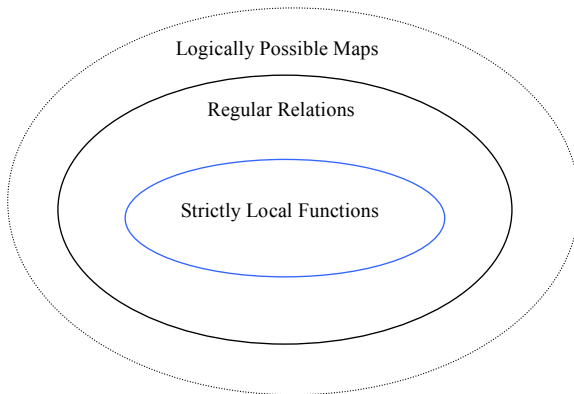


Riggle (2004); Gerdemann & Hulden (2012); Buccola 2013

Phonological maps are Regular Relations

- Not the best characterization.
 - Regular relations can also describe rules that are unattested and phonologically 'odd'.
- Doesn't help with learning.
 - The class of regular relations is not learnable from positive data.

Phonological maps are Strictly Local Functions



Chandlee (2014)

Strictly Local Functions

- Input Strictly Local Functions (ISL)
- Output Strictly Local Functions (OSL)

Input Strictly Local Functions

- An ISL function is actually ISL- k for some integer k .
- The output for a given input can be determined by only paying attention to k contiguous segments at a time (i.e., bounded memory).
- For a phonological map, k = the length of the target + the triggering context.

Input Strictly Local Functions

(1) Nasal place assimilation

a. $f(iN + \text{possible}) = \text{impossible}$

$$k = |Np| = 2$$

Input Strictly Local Functions

i N p o s s i b l e

Input Strictly Local Functions

i N p o s s i b l e #
i

Input Strictly Local Functions

i N p o s s i b l e #
i

Input Strictly Local Functions

i N p o s s i b l e #
i mp

Input Strictly Local Functions

i N p o s s i b l e #
i mp o

Input Strictly Local Functions

i N p o s s i b l e #
i mp o s

ISL: Typological Prediction

- A significant range of phonological processes (local substitution, deletion, epenthesis, synchronic metathesis) can be modeled with ISL functions.
- Approximately 95% of the processes in P-Base (v.1.95, Mielke (2008)) meet this criterion (i.e., are ISL).
- Chandlee, Jardine, Heinz (2014) also show that the opaque maps discussed in Baković (2007) are ISL.

ISL: Learning Result

The class of ISL functions is efficiently identifiable in the limit from positive data (Chandlee, Eyraud, Heinz 2014) in the sense of de la Higuera (1997).

Non-ISL processes

Because ISL functions only pay attention to the input, the trigger of the process must be present underlyingly.

Nasal spreading

(2) Johore Malay (Onn 1980)

- a. $[+nasal][-cons] \mapsto [+nasal][+nasal]$
- b. $pəŋawasan \mapsto pəŋãwãsan$, 'supervision'

Output Strictly Local Functions

p ə ŋ a w a s a n #
p

Output Strictly Local Functions

p ə ŋ a w a s a n #
p ə

Output Strictly Local Functions

p ə ɲ a w a s a n #
p ə ɲ

Output Strictly Local Functions

p ə ŋ a w a s a n #
p ə ŋ ã

Output Strictly Local Functions

#	p	ə	ŋ	a	w	a	s	a	n	#
#	p	ə	ŋ	ã	ĩ					

Output Strictly Local Functions

#	p	ə	ŋ	a	w	a	s	a	n	#
#	p	ə	ŋ	ã	ĩ	ã				

Spreading versus Harmony

- Though a target of the assimilation can be arbitrarily far from the trigger, the iterative nature of the spreading makes it a local process.
- This contrasts with harmony processes in which intervening 'potential' targets are skipped (Hansson 2001, Rose & Walker 2004).

Transparent Vowels

- (3) Finnish front-back harmony (Goldsmith 1995, van der Hulst & van de Weijer 1995)
- a. værtt*i*næ 'spinning wheel'
 - b. værttinæ-llæ-*n*i-hæn 'with spinning wheel, as you know'
 - c. tuoli 'chair'
 - d. tuoli-lla 'on the chair'

Transparent Vowels

værttinæ-llA-nl-hAn



✓ OSL

værttinæ-llæ-nw-hæn



✓ OSL

værttinæ-llæ-ni-hæn

(Clements 1977; Vago 1980)

Transparent Vowels

- Assuming harmony is local (Gafos 1996, Ní Chiosáin & Padgett 1997), then transparent vowels create opaque maps:
 - Underapplication: high vowels don't harmonize
 - Overapplication: non-high vowels that follow non-harmonizing high vowels *do* harmonize

Transparent Vowels

Targeted Constraints analysis (Baković & Wilson 2000)

(4) Wolof ATR harmony (Archangeli & Pulleyblank 1994)

- a. reer-oon 'was lost'
- b. rɛɛr-ɔɔn 'had dinner'
- c. tɛɛr-**u**w-ɔɔn 'welcomed'

Transparent Vowels

Targeted Constraints analysis (Baković & Wilson 2000)

$t\epsilon\epsilon r-Uw-OO_n \mapsto t\epsilon\epsilon r-uw-ɔɔ_n$

AGREE(ATR) prefers the full assimilation candidate $t\epsilon\epsilon r-uw-ɔɔ_n$, but the targeted constraint $\rightarrow NO(+HI, -ATR)$ selects the candidate that differs minimally from full assimilation (i.e., the transparency candidate): $t\epsilon\epsilon r-uw-ɔɔ_n$.

Long-distance Assimilation

- (5) Kikongo (Rose and Walker 2004)
- a. tu-kun-idi \mapsto tu-kun-ini 'we planted'
 - b. tu-nik-idi \mapsto tu-nik-ini 'we ground'

Long-distance Assimilation

tu-nik-idi

✓ OSL

tu-nĩk-ĩnĩ

✓ OSL

tu-nik-ini

Interim Summary

harmony \leftarrow OSL \circ OSL
 \downarrow
spreading \leftarrow OSL
 \downarrow
adjacency \leftarrow ISL

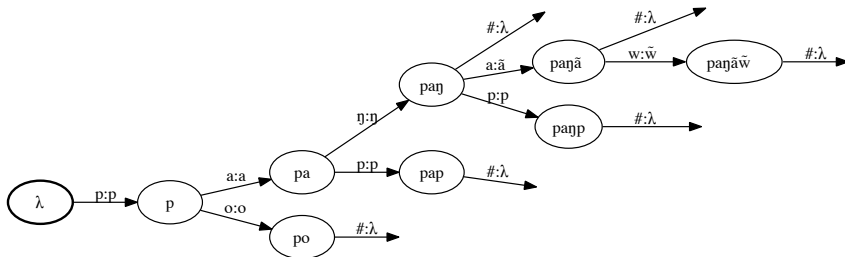
Learning OSL

- Modification to the ISLFL Algorithm already proven to learn the class of ISL functions (Chandlee, Eyraud, Heinz 2014).

Learning OSL

- Starts with a finite state representation of the observed data.
- $\{(pa\eta, pa\eta), (pa\eta a, pa\eta \tilde{a}), (pa\eta aw, pa\eta \tilde{a}\tilde{w}), (pa\eta s, pa\eta s), (pap, pap), (po, po) \dots \}$

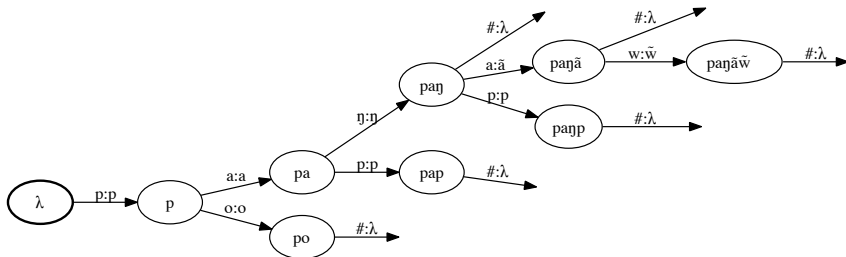
Learning OSL



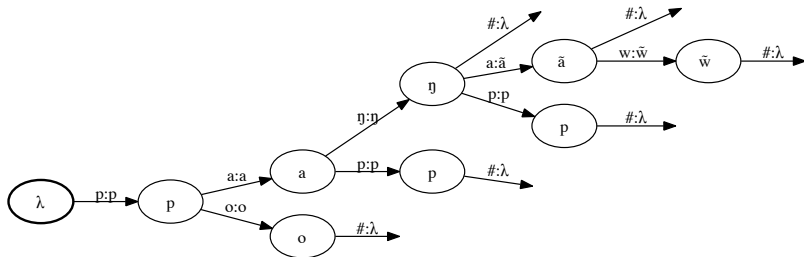
Learning OSL

- Generalize to unobserved inputs by merging states.
- State merging criterion reflects the defining property of the OSL class (i.e., only need to remember the ‘most recent’ output symbols).

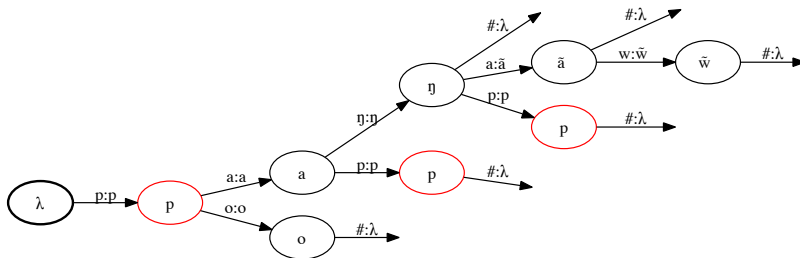
Learning OSL



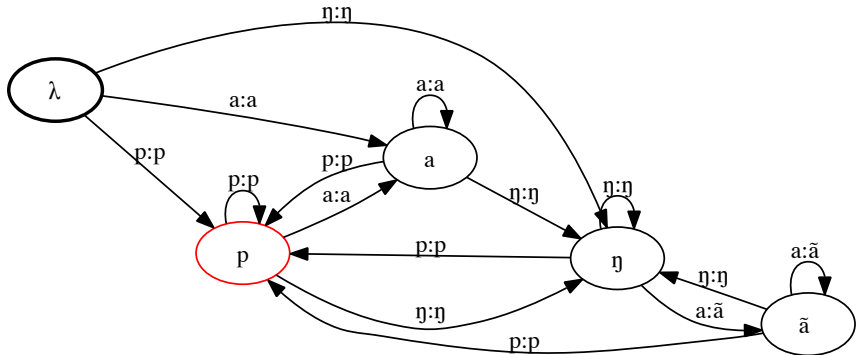
Learning OSL



Learning OSL



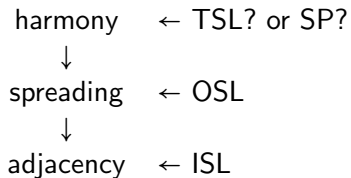
Learning OSL



Future Work

- What do the intermediate forms look like?
 - how abstract can they be?
- How is the composition learned?
 - Are the segments that don't surface still available within the 'feature space'?

Future Work



Conclusions

- Proposal for representing and learning long-distance processes as well as an approach to the opacity problem.
- Progress on all goals for a theory of the phonological grammar: sufficiently expressive, maximally restrictive, and learnable.

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- ◀ ◻ ▶ ◀ ◻ ▶ ◀ ≡ ▶ ◀ ≡ ▶ ≡

Schwa deletion

(6) French (Dell 1973, 1980, Noske 1993)

a. $VCəCV \mapsto VCCV$

(7) tu devenais, 'you became'

a. $ty\ dəvənɛ \mapsto ty\ dvənɛ$

b. $ty\ dəvənɛ \mapsto ty\ dəvnɛ$

c. $ty\ dəvənɛ \mapsto *ty\ dvnɛ$

Strict Locality

Post-nasal obstruent voicing: Valid surface strings are a SL-2 language.

TVNDV in the language.

NVNTV in the language.

TVNTV *necessarily* in the language.

Optimization

IDENT, DEP \gg **ab* \gg MAX

Non-regular relation:

$$a^n b^m \mapsto a^n, \text{ if } m < n$$

$$a^n b^m \mapsto b^m, \text{ if } n < m$$