### The Computational Nature of Phonological Generalizations

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Overview

#### In this talk...

- 1. Explain why computational characterizations of language patterns matter.
- 2. Explain the subregular computational classes that phonological generalizations appear to belong to.
- 3. Provide some pyscholinguistic evidence that the boundaries of these computational classes are psychologically real.

- Prof. Jim Rogers (Earlham College)
- Prof. Herbert G. Tanner (UD)
- Prof. Bill Idsardi (UMCP)
- Dr. Regine YeeKing Lai, PhD 2012
- Cesar Koirala (PhD exp. 2013)
- Jane Chandlee (PhD exp. 2014)
- Adam Jardine (PhD exp. 2016)
- Amanda Payne (PhD exp. 2016)
- Huan Luo (PhD exp. 2017)
- Brian Gainor (LDC)





Bert



Regine



Cesar



Adam



Amanda

Unpictured Bill, Jane, Huan, Brian

### What are phonological generalizations?

- 1. Phonotactics
- 2. Phonological processes; i.e. mappings
- 3. Contrast (not part of this talk)

### Phonotactics - Knowledge of word well-formedness

ptak thole hlad plast sram mgla vlas flitch dnom rtut

Halle, M. 1978. In Linguistic Theory and Pyschological Reality. MIT Press.

possible English words	impossible English words
thole	$\operatorname{ptak}$
plast	hlad
$\operatorname{flitch}$	$\operatorname{sram}$
	mgla
	vlas
	$\operatorname{dnom}$
	$\operatorname{rtut}$

#### Question

Overview

How do English speakers know which of these words belong to different columns?

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	m mgla
	$ ext{vlas}$
	<u>dn</u> om
	$\operatorname{rtut}$

#### Question

How do English speakers know which of these words belong to different columns?

#### Example

### Example

All possible English words are in the set; all impossible words are out of the set.

mql

### Example

$$mgl \cdot \Sigma^*$$

#### Example

$$\overline{mgl\cdot \Sigma^*}$$

#### Example

$$\overline{mgl \cdot \Sigma^*} \cap \overline{pt \cdot \Sigma^*} \cap \dots$$

#### Example

A phonotactic constraint in Yawelmani Yokuts prohibits sequences of three consonants (\*CCC).

All logically possible strings with no CCC sequence are in the set; all others with at least one CCC sequence are out of the set.

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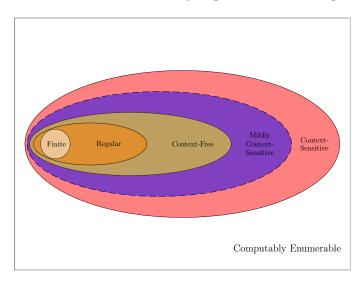
$$\overline{\Sigma^* \cdot CCC \cdot \Sigma^*}$$

#### Example

Any markedness constraint in Optimality Theory.

All surface forms with zero violations are in the set; all surface forms with nonzero violations are out of the set.

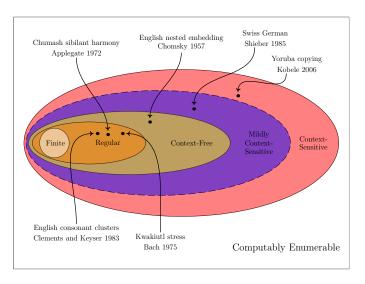
# Classifying Sets of Strings



computably enumerable contextsensitive mildly contextsensitive context-free regular finite

Figure: The Chomsky hierarchy

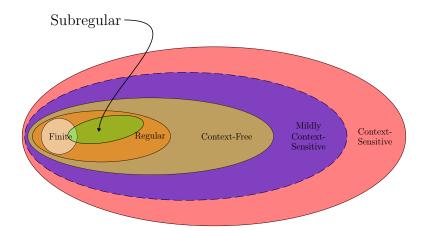
### Classifying Sets of Strings



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Overview

### What is subregular?



# There is room at the bottom

### Better characterizations of phonological patterns

- Leads to stronger universals
- Leads to new hypotheses regarding what a humanly possible phonological pattern is

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• Are the stronger universals useful for learning?

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### Better characterizations of phonological patterns

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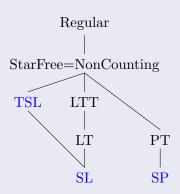
### Payoffs for better understanding learning

• Are the stronger universals useful for learning?

#### Payoffs for natural language processing

- Insights can be incorporated into NLP algorithms
- Factoring and composition may occur with lower complexity

### Interesting subregular classes of stringsets



(McNaughton and Papert 1971, Rogers et al. 2010, 2012, Heinz et al. 2011)

LTTLocally Threshold Testable LTLocally Testable SLStrictly Local

TSLTier-based Strictly Local Piecewise Testable PTSPStrictly Piecewise

### Phonotactics - Knowledge of word well-formedness Samala Version

[tojonowonowa] stojonowonowaſ stojonowonowas [tojonowonowas pisotonosikiwat pisotonosikiwat sanisotonosikiwas fanipisotono fikiwas

# Phonotactics - Knowledge of word well-formedness Samala Version

possible Samala words	impossible Samala words
∫tojonowonowa∫	$\operatorname{stojonowonowa}$
stojonowonowas	∫tojonowonowas
pistonoskiwat	pisotono∫ikiwat
sanisotonoskiwas	∫anipisotono∫ikiwas

- 1. Question: How do Samala speakers know which of these words belong to different columns?
- 2. By the way, ftoyonowonowaf means 'it stood upright' (Applegate 1972)

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possible Samala words	impossible Samala words
∫tojonowonowa∫	stojonowonowa∫
stojonowonowa <mark>s</mark>	∫tojonowonowas
pi <mark>s</mark> tono <mark>s</mark> kiwat	pi <mark>s</mark> otono∫ikiwat
sanisotonoskiwas	∫anipi <mark>s</mark> otono∫ikiwa <mark>s</mark>

- 1. Question: How do Samala speakers know which of these words belong to different columns?
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# Phonotactics - Knowledge of word well-formedness Language X

possible words of Language X	impossible words of Language X
∫otko∫	$\mathrm{sotko} \!\! \int$
∫o∫ko∫	$\int\!\! { m otkos}$
∫osoko∫	∫o∫kos
so∫okos	$\operatorname{sosko} \int$
sokosos	
pitkol	
pisol	
pi∫ol	

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pi∫ol	

Sibilant sounds which begin and end words must agree (but not ones word medially).

possible words of Language Y	impossible words of Language Y
∫otko∫	∫o∫ko∫
sotkof	∫osko∫
$\int \!\! \mathrm{otkos}$	so∫kos
pitkol	∫o∫kos
sofkostof	sosko∫
	soksos
	piskol
	pi∫kol

# Phonotactics - Knowledge of word well-formedness Language Y

possible words of Language Y	impossible words of Language Y
∫otko∫	∫o∫ko∫
sotko∫	∫osko∫
∫otkos	so∫kos
pitkol	∫o∫kos
so∫kosto∫	sosko∫
	soksos
	pi <mark>s</mark> kol
	pi <mark>∫</mark> kol

Words must have an *even number* of sibilant sounds.

# Typology

#### Attested Phonotactic Patterns

- 1. Words don't begin with mgl. (English)
- 2. Words don't contain both f and s. (Samala)

#### Unattested Phonotactic Patterns

- 1. Words don't begin and end with disagreeing sibilants. (Language X = First/Last Harmony)
- 2. Words don't contain an even number of sibilants. (Language Y = \*EVEN-Sibilants)

### What's the explanation?

#### Optimality Theory

- 1. Constraints like \*#mgl and \*[+strident, $\alpha$  anterior]...[+strident, $-\alpha$  anterior] are part of CON.
- 2. Constraints like \*EVEN-Sibilants or \*#[+strident, $\alpha$  anterior]...[+strident, $-\alpha$  anterior]# are not.

### Phonetically-based Phonology

- 1. There are perceptual and/or articulatory reasons for constraints like \*#mgl and
  - \*[+strident, $\alpha$  anterior]...[+strident, $-\alpha$  anterior].
- 2. There are no such reasons for constraints like
  - \*EVEN-Sibilants or
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What are those reasons?

# First/Last Harmony

- 1. Long-distance assimilation is well-attested (Hansson, 2001; Rose & Walker, 2004)
- 2. Word edges in phonology are privileged positions (Beckman, 1997; Endress, Nespor & Mehler, 2009; Fougeron & Keating, 1997).

#### Question

What theory of perception or articulation prevents there from being harmony only in privileged positions?

### First/Last Harmony

### Are the memory requirements greater?

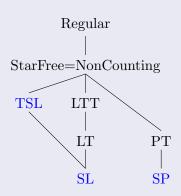
Given the pattern templates, the answer seems to be no.



#### \*EVEN-Sibilants

- It's plausible to me at least that perception or articulation should be able to explain the absence of counting mod npatterns in phonology, but I haven't seen any explicit connection.
- Whatever it is, it *should* connect to the computational properties discussed here.

## A computational explanation



- 1. Constraints like \*#mgl are Strictly Local.
- 2. Constraints like \*[+strident, $\alpha$  anterior]...[+strident, $-\alpha$  anterior] are Strictly Piecewise.
- 3. Constraints like First Last Harmony are Locally Testable.
- 4. Constraints like \*EVEN-Sibilants are Counting (properly regular).

## Strictly Local Stringsets

#### Substrings

String u is a substring of w iff  $w \in \Sigma^* u \Sigma^*$ . We write  $u \leq w$ .

$$F_k(w) = \begin{cases} \{u \in \Sigma^* \mid u \leq w \land |u| = k\} & |w| \geq k \\ \{w\} & \text{otherwise} \end{cases}$$

Example:  $\alpha = abbcac$ ;  $F_2(\alpha) = \{ab, bb, bc, ca, ac\}$ .

#### Strictly Local (SL) Stringsets

$$L \in SL_k \iff \left(\exists G \subseteq F_k(\rtimes \cdot \Sigma^* \cdot \ltimes)\right)$$
$$\left[L = L(G) = \left\{w \in \Sigma^* \mid (\forall u \in G) \ [u \not\preceq \rtimes \cdot w \cdot \ltimes]\right\}\right]$$

Example: If  $G = \{ac, ad\}$  then  $\alpha \notin L(G)$ .

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## Strictly Piecewise Stringsets

#### Subsequences

String  $u = \sigma_1 \cdots \sigma_n$  is a subsequence of w iff  $w \in \Sigma^* \sigma_1 \Sigma^* \cdots \Sigma^* \sigma_n \Sigma^*$ . We write  $u \sqsubseteq w$ .

$$P_k(w) = \left\{ u \in \Sigma^* \mid u \sqsubseteq w \land |u| \le k \right\}$$

Example:  $\alpha = abcd$ ;  $P_2(\alpha) = \{\lambda, a, b, c, d, ab, ac, ad, bc, bd, cd\}$ .

### Strictly Piecewise (SP) Stringsets

$$L \in SP_k \iff \left(\exists G \subseteq P_k(\Sigma^*)\right)$$
$$\left[L = L(G) = \left\{w \in \Sigma^* \mid (\forall u \in G) \ [u \not\sqsubseteq w]\right\}\right]$$

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Example: If  $G = \{aa, ad\}$  then  $\alpha \notin L(G)$ .

## Locally Testable and Piecewise Testable Stringsets

### Locally Testable (LT) Stringsets

These are finite boolean combinations of SL languages. It is also known:

$$L \in LT_k \iff \forall w, v \in \Sigma^* \left[ F_k(w) = F_k(v) \Rightarrow \left[ w \in L \Leftrightarrow v \in L \right] \right]$$

#### Piecewise Testable (PT) Stringsets

These are finite boolean combinations of SP languages. It is also known:

$$L \in PT_k \iff \forall w, v \in \Sigma^* \left[ P_k(w) = P_k(v) \Rightarrow \left[ w \in L \Leftrightarrow v \in L \right] \right]$$

## Tiers: Ignoring inconsequential events

#### Tiers

A tier T is a subset of  $\Sigma$ .

#### Definition

The erasing (projection) function:

$$E_T(\sigma_1 \cdots \sigma_n) = \tau_1 \cdots \tau_n$$

where  $\tau_i = \sigma_i$  iff  $\sigma_i \in T$  and  $\tau_i = \lambda$  otherwise

#### Example

If 
$$\Sigma = \{a, b, c\}$$
 and  $T = \{b, c\}$  then

$$E_T(aabaaacaaabaa) = bcb$$

## Typology of phonotactic patterns

#### Phonotactic Patterns derived from

- ✓ Adjacency constraints are SL
- Consonantal harmony are SP/TSL
- Consonantal disharmony are TSL
- ✓ Vowel harmony without neutral vowels are SP/TSL
- ✓ Vowel harmony with opaque vowels are TSL
- ✓ Vowel harmony with transparent vowels are SP/TSL
- ✓ Stress patterns are ... (attend Jim's talk next week!)

Heinz 2007, 2010, Rogers et al. 2010, Heinz et al. 2011

1.  $SL_k$ ,  $SP_k$ , and  $TSL_{T,k}$  are provably identifiable in the limit from positive data by an incremental, set-driven, polytime learning algorithms.

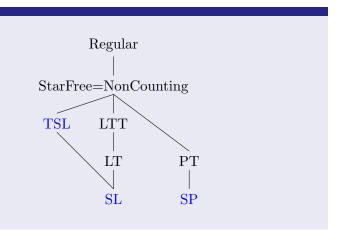
Garcia et al. 1991, Heinz 2007, 2010, Rogers et al. 2010

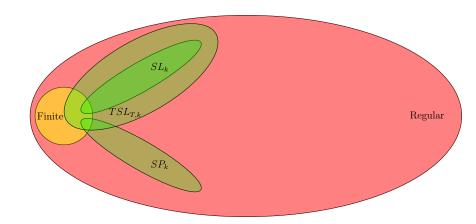
Heinz et al. 2011, Heinz et al. 2012

- k (and T) must be known a priori.
- k appears to be small for phonology (perhaps  $\leq 5$ ).
- 2. Provably-correct stochastic versions of these algorithms exist which learn probability distributions over stringsets.

Jurafsky and Martin 2008, Heinz and Rogers 2010

3. Phonological features and syllables can be fully integrated into these algorithms without compromising their correctness. Heinz and Koirala 2010, Koirala et seq.





## Can college students learn First Last Harmony?

Psychological reality

#### Artificial language learning experiments

- 1. Subjects are exposed to training items (exemplars of a pattern).
- 2. Subjects are tested on novel items, some which exemplify the target pattern, and some which don't.
  - Which word do you think more likely belongs to the language you just heard?

# Comparative artificial language learning experiments (Lai 2012, under review)

		Pattern Type	
		SL/SP/TSL (SH)	non-SL/SP/TSL (FL)
0	1	Learnable	Learnable
	2	Unlearnable	Unlearnable
Outcomes	3	Learnable	Unlearnable
	4	Unlearnable	Learnable

- It is not possible to test for the *unlearnability* of some pattern.
- Instead, Lai (2012) tests the *comparative* learnability.

Psychological reality

## Subjects

66 adult native English speakers

#### All Stimuli

Training and test items were C<sub>1</sub>V.C<sub>2</sub>V.C<sub>3</sub>VC<sub>4</sub> (tryisyllabic), containing 3 sibilants.

•  $C_1 \& C_4$ : sibilants

• C<sub>2</sub> & C<sub>3</sub>: either sibilant or [k]

### Training

 $40 \text{ words} \times 5 \text{ repetitions} = 200 \text{ words}$ . Subjects listened and repeated each word. 3 Training Conditions:

SH: 
$$[s...s...s], [\int...\int...]$$

FL: 
$$[s...s...s]$$
,  $[[...]...]$ ,  $[s...]...s]$ ,  $[[...s...]]$ 

Control: No training

## Testing (Lai 2012, under review)

Psychological reality

#### Two alternative forced choice

Words were presented in pairs (minimally different)

E.g. [sakisis] vs. [fakisis]

- In the FL and SH conditions, subjects had to answer "Which word do you think belongs to the language you just heard?"
- In the control condition, they were asked "Which word do you prefer?"
- 48 pairs in total

# Stimuli (Lai 2012, under review)

Psychological reality

## Three Stimuli Types

FL/SH	[sokosos]
*FL/*SH	[sokoso∫]
FL/*SH	[soko fos, fokoso f]

- These 3 types of stimuli were pitted against each other and generated 3 types of pairings.
  - (a) FL/\*SH vs. \*FL/\*SH (also includes \*FL/\*SH vs. FL/\*SH)
  - (b) FL/SH vs. \*FL/\*SH (also includes \*FL/\*SH vs. FL/SH)
  - (c) FL/\*SH vs. FL/SH (also includes FL/SH vs. FL/\*SH)
- The order of presentation was counter-balanced across types

# Data Analysis (Lai 2012, under review)

The dependent variable for each pairing is different, so they were analyzed separately

- (a) FL/\*SH vs. \*FL/\*SH Rate of choosing FL/\*SH
- (b) FL/SH vs. \*FL/\*SH Rate of choosing FL/SH
- (c) FL/\*SH vs. FL/SH Rate of choosing FL/SH

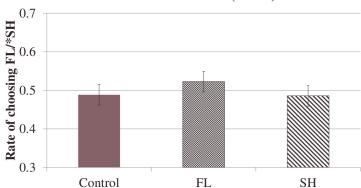
## Predictions (Lai 2012, under review)

If subjects internalized the pattern they were exposed to during training, they should perform as follows.

	Pairs		
	FL/*SH vs.	FL/SH vs.	FL/SH vs.
Conditions	*FL/*SH	*FL/*SH	FL/*SH
SH	No preference	FL/SH	FL/SH
FL	FL/*SH	FL/SH	No preference
Control	No preference	No preference	No preference

Psychological reality





Control

0.7

0.6

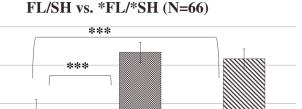
0.5

0.4

0.3

Rate of choosing FL/SH

## Results (Lai 2012, under review)

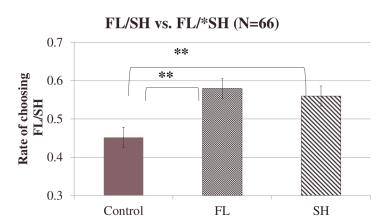


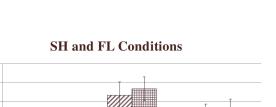
FL

SH

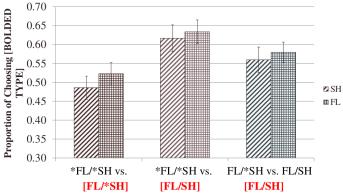
## Results (Lai 2012, under review)

Psychological reality





Psychological reality



- 1. Subjects in the SH condition behaved as if they had internalized the SH pattern.
- 2. Subjects in the FL condition behaved as if they had internalized the SH pattern, *not* the FL pattern!

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## Discussion (Lai 2012, under review)

- 1. Subjects in the SH condition behaved as if they had internalized the SH pattern.
- 2. Subjects in the FL condition behaved as if they had internalized the SH pattern, not the FL pattern!

#### Conclusion

The heavy bias for SH can be understood if only phonotactic patterns which can be modeled as SL, SP, or TSL stringsets are the humanly learnable ones.

# Explaining Allomorphy

Lithuanian Verbal Prefixes exhibit allomorphy.

at-eiti	'to arrive'
at-imti	'to take away'
at-ne∫ti	'to bring'
at-leisti	'to forgive'
at-likti	'to complete'
at-korpti	'to rise'
at-pra∫i:ti	'to ask'
at-kurti	'to reestablish'
ad-bekti	'to run up'
ad-gauti	'to get back'
ad-bukti	'to become blunt'
ad-gimti	'to be born again'

#### Lithuanian Verbal Prefixes exhibit allomorphy.

ap-eiti	to circumvent
ap-ieſkoːti	to search everywhere
ap-akti	to become blind
ap-moːkiːti	to train
ap-temdiːti	to obscure
ap-∫aukti	to proclaim
ab-gauti	to deceive
ab-z <sup>j</sup> ureti	to have a look at
ab-zelti	to become overgrown
ab-dauzi:ti	to damage
ab-draski:ti	to tear

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1 1	
ab-zelti	to become overgrown
ab-zelti ab-dauzizti	to become overgrown to damage
9	_

• Phonology is the theory that posits a single lexical representation /ap/ and a mapping which maps a /p/ which is immediately followed by a voiced obstruent to [b].

# This knowledge can be represented with string relations

## Example

In Lithuanian, voiceless stops become voiced when immediately followed by voiced obstruents. Every pair (u, w) such that u maps to w is in the set and every other pair is out.

```
ateiti
                    ateiti
atbekti
                    adbekti
              \mapsto
apmozkizti
                    apmozkizti
               \mapsto
apgauti
                    abgauti
               \mapsto
```

- 1. Optional, left-to-right, right-to-left, and simultaneous application of SPE-style rules  $A \longrightarrow B / C \_D$  (where A,B,C,D are regular expressions) describe regular relations, provided the rule cannot reapply to the locus of its structural change.

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- 3. Regular relations are closed under composition.

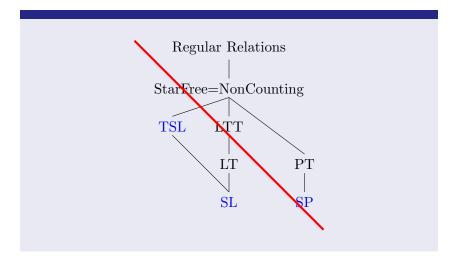
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- 2. Rule ordering is functional composition (finite-state transducer composition).
- 3. Regular relations are closed under composition.
- 4. SPE grammars (finitely many ordered rewrite rules of the above type) can describe virtually all phonological patterns.

- 1. Optional, left-to-right, right-to-left, and simultaneous application of SPE-style rules  $A \longrightarrow B / C \_D$  (where A,B,C,D are regular expressions) describe regular relations, provided the rule cannot reapply to the locus of its structural change.
- 2. Rule ordering is functional composition (finite-state transducer composition).
- 3. Regular relations are closed under composition.
- 4. SPE grammars (finitely many ordered rewrite rules of the above type) can describe virtually all phonological patterns.

## Therefore, phonological mappings are regular relations.

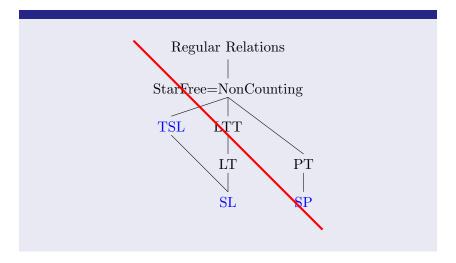
Regardless of whether they are described with SPE or OT grammars.

## Regular sets $\neq$ Regular relations

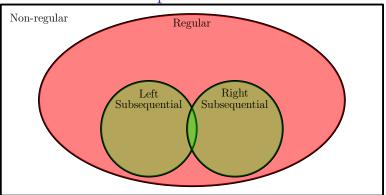


There are no similar subregular hierarchies for relations

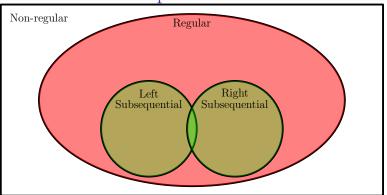
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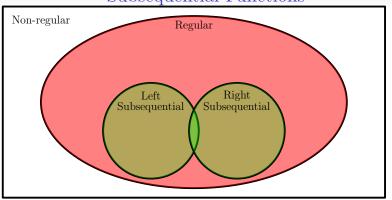
There are no similar subregular hierarchies for relations (yet)



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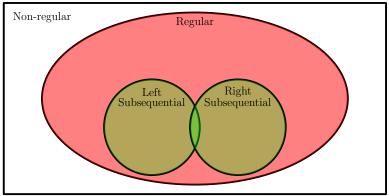
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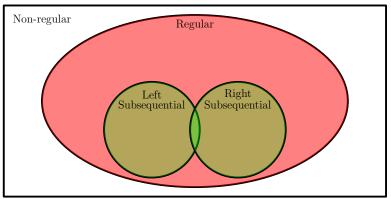
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- 6. The long-distance consonantal harmony patterns in Hansson (2001) are left or right subsequential (Luo, 2013 MS).

# What is not subsequential?



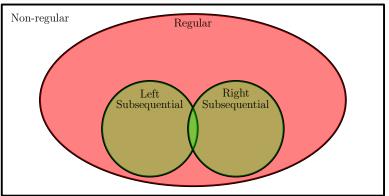
7. The "Majority Rules" vowel harmony pattern is not regular and the "Sour Grapes" vowel harmony pattern is neither left nor right subsequential (Heinz and Lai, 2012 MS).

## Suprasegmental phonology



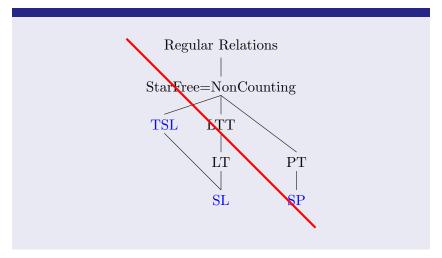
- 8. Unbounded Tone Plateauing is neither left nor right subsequential (Jardine, 2012 MS).
- Paraphasing Yip (2001) and Hyman (2011): "Tone can do everything segmental phonology can do and more!"

#### The future...



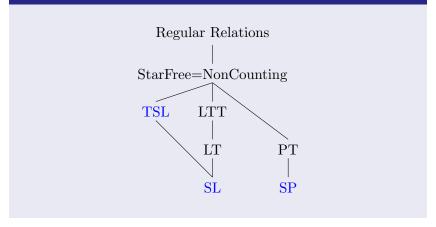
- There are stronger properties than "being left or right subsequential".
- Chandlee's in progress thesis (exp. 2014) aims to define Strictly Local mappings to capture local phonological processes, including many of the ones mentioned above.
- Chandlee and Koirala (2013, PLC) present the first learning results following this line of research.

## Regular sets $\neq$ Regular relations



There are no similar subregular hierarchies for relations

### Regular sets may inform Regular relations



There are no similar subregular hierarchies for relations (yet)

#### Conclusions

- 1. Computational analysis of stringsets and string mappings (ongoing) is yielding natural classes of pattern complexity.
- 2. When phonological patterns are studied through this lens, strong computational properties are revealed, which:
  - 2.1 appear to make the right kind of cuts between attested and unattested patterns.
  - 2.2 appear to draw the right distinctions between segmental and suprasegmental phenomenon.
  - 2.3 are strong enough to make learning possible from positive evidence.
  - 2.4 make experimentally testable predictions.

#### THANK YOU

