

Mathematical Linguistics & Typological Complexity

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(Some) Big Questions

- ► Are there **laws** that govern linguistic knowledge?
- ▶ Why are those the laws?
- Do they relate to typological gaps, i.e. logically possible patterns we don't (seem to) find?
- What can we infer about human learning processes?

Cross-disciplinarity for the win

- Stand on the shoulders of giants.
- Cross-fertilization and multiple explanatory levels.
- Yields new generalizations and data.

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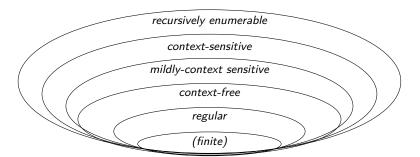
- Stand on the shoulders of giants.
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Outline

- 1 Linguistics and Formal Language Theory
- 2 Refining the Hierarchy via Typological Insights
- 3 Artificial Grammar Learning
- 4 Summing Up & Future Directions

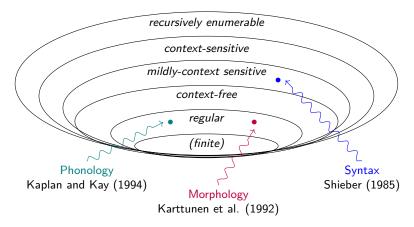
Computational Theories of Language

Languages (stringsets) can be classified according to the complexity of the grammars that generate them.



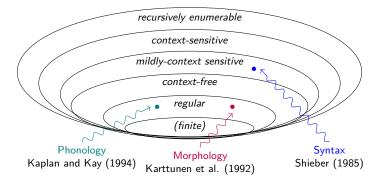
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Subregular Classes Cross-Fertilization Artificial Grammar Learning Conclusion

Precise Theories ⇒ Precise Predictions



Precise predictions for:

- ightharpoonup typology ightarrow e.g. no center embedding in phonology
- ightharpoonup learnability ightarrow e.g. no Gold learning for regular languages
- ightharpoonup cognition ightarrow e.g. finitely bounded working memory

Context-sensitive

Classifying Patterns

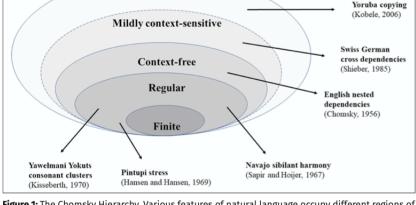
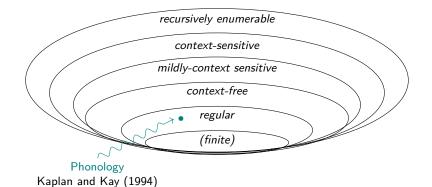


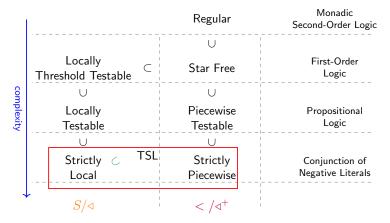
Figure 1: The Chomsky Hierarchy. Various features of natural language occupy different regions of the hierarchy. Figure reproduced from Figure 1 in Heinz (2010: 634) with permission.

Subregular Classes Cross-Fertilization Artificial Grammar Learning Conclusion

Phonology as a Regular System

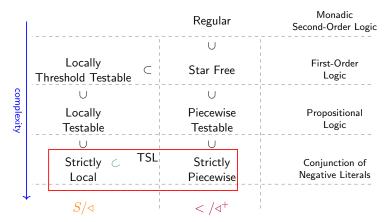


Beyond Monolithic Classes: Subregular Languages



- Multiple equivalent characterizations:
 - ⇒ algebraic. logic. automata...

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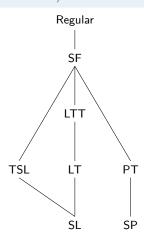
Phonology as a Subregular System

Subregular Phonotactics

► Majority of phonological patterns are subregular (Heinz 2011a,b; Chandlee 2014; Graf 2017:a.o.).

Most phonological and morphological rules correspond to p-subsequential relations.

(Mohri 1997)



Phonology as a Subregular System

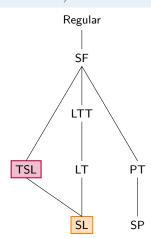
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A caveat: Mostly phonotactics today!



Local Dependencies in Phonology

Word-final devoicing

Forbid voiced segments at the end of a word

- (1) a. * rad
 - b. rat

Intervocalic voicing

Forbid voiceless segments in between two vowels

- (2) a. * faser
 - b. fazer

These patters can be described by **strictly local** (SL) constraints.

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Local Dependencies in Phonology are SL

Example: Word-final devoicing

- Forbid voiced segments at the end of a word: *[+voice]\$
- **German**: *z\$, *v\$, *d\$ (\$ = word edge).

Example: Intervocalic voicing

- Forbid voicess segments in-between two vowels: *V[-voice]V
- German: *ase, *ise, *ese, *isi, ...
 - **\$** faser**\$**

\$ fazer\$

Local Dependencies in Phonology are SL

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```
$ faser$
```

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- **Compane Fig. 3 ► Compane Proof Compane Proof Compane Proof Compane Compane**

Example: Intervocalic voicing

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- ➤ Samala Sibilant Harmony
 Sibilants must not disagree in anteriority.
 (Applegate 1972)
 - (3) a. * hasxintilawaſ
 - b. * ha∫xintilawa**s**
 - c. ha∫xintilawa∫

Example: Samala

```
*$hasxintilawa∫$
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```
*$ has xintilawas$
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 - (3) a. * hasxintilawa
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Example: Samala

But: Sibilants can be arbitrarily far away from each other!

```
*$stajanowonwa∫$
```

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Example: Samala

```
*$ has xintilawa \s\$
$ ha\s\xintilawa \s\$
```

▶ But: Sibilants can be arbitrarily far away from each other!

```
*$<mark>s</mark>tajanowonwa∫$
```

Locality Over Tiers

```
*$<mark>s</mark>tajanowonwa∫$
```

- ► Sibilants can be arbitrarily far away from each other!
- **Problem**: SL limited to locality domains of size *n*;

Tier-based Strictly Local (TSL) Grammars (Heinz et al. 2011)

- Projection of selected segments on a tier T (Goldsmith 1976)
- Strictly local constraints over T determine wellformedness
- Unbounded dependencies are local over tiers



Locality Over Tiers

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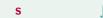
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Unbounded Dependencies are TSL

- Let's revisit Samala Sibilant Harmony
 - (4) a. * hasxintilawa
 - b. * ha∫xintilawas
 - c. haʃxintilawaʃ
- ► What do we need to project? [+strident]
- ► What do we need to ban? *[+ant][-ant],*[-ant][+ant] I.E. *sʃ, *sʒ, *zʃ, *zʒ, *ʃs, *ʒs, *ʃz, *ʒz

Example: TSL Samala



* \$hasxintilaw[s

ok\$ha∬xintilaw∬\$

Unbounded Dependencies are TSL

- Let's revisit Samala Sibilant Harmony
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Interim Summary: SL and TSL for Phonology

- ► Linguistically natural (Goldsmith 1976)
- ► Captures wide range of phonotactic dependencies (McMullin 2016)
- ► Provably correct and efficient learning algorithms (Jardine and McMullin 2017)
- ► Rules out unattested patterns (cf. Lai 2015, Aksenova et al. 2016, Graf & De Santo 2019, a.o.)

But:

- ► Typological variation is complex, knowledge is limited
- ► Can we truly gain cognitive insights?

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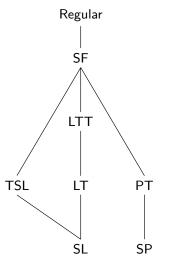
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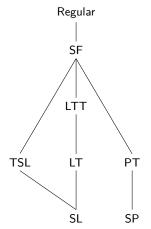
► But not every long-distance pattern is TSL! (McMullin 2016, Mayer & Major 2018, De Santo & Graf 2019)

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Concurrent Processes (De Santo and Graf, 2019)

Observation

 TSL is not closed under intersection



- We want to also account for multiple processes
 So we can cover the complete phonotactics of a language
- Multiple non-interacting processes in attested patterns

A TSL Outlier

Sibilant Harmony in IMDLAWN TASHLHIYT (McMullin2016)

1) Underlying causative prefix /s(x)-/

Base Causative

"be evacuated" uga s:-uga a.

b. asitwa s-asitwa "settle, be levelled"

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2) Sibilant harmony

Base Causative

fia[r [- fia[r "be full of straw, of discord" a.

h. "be sold" zː-nza nza

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        Causative
  fiaſr ſ- fiaſr
                      "be full of straw, of discord"
a.
h.
                      "be sold"
   nza
           z:-nza
3) Sibilant voicing harmony blocked
    Base
            Causative
a. ukz sz-ukz "recognize"
   qːuʒːi ∫- quʒːi "be dislocated, broken"
b.
```

Generalization (1/2)

Sibilants must agree in anteriority and voicing.

Grammar

$$T = \{ g, s, z, \}$$

$$S = \{ *sg, *sz, *sf, *gs, *fs, *zs, *zf, *zg, *fz, *fg, *gf, *gz \}$$

* z m: 3 d a w |

ok 3 m: 3 d a w |

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z 3

* z m: 3 d a w l

 ok 3 m: 3 d a w l

Generalization (1/2)

Sibilants must agree in anteriority and voicing.

Grammar

$$T = \{ g, s, z, j \}$$

$$S = \{ *sg, *sz, *sj, *gs, *js, *zs, *zj, *zg, *jz, *jg, *gj, *gz \}$$

* z m: 3 d a w l

 ok 3 m: 3 d a w [

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* z m: 3 d a w |

3 3

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* z m: 3 d a w l

ok 3 m: 3 d a w I

Generalization (2/2)

Voiceless obstruents block agreement in voicing.

$$T = \{ \text{ $\tt x$, $\tt s$, $\tt z$,, $\tt q} \}$$

$$S = \{ \text{ $\tt *s$, $\tt *s$, $\tt *s$, $\tt *g$, $\tt *f$, $\tt *z$, $\tt *z$, $\tt *z$, $\tt *f$, $\tt *g$, $\tt$$

ok
 \int q u \mathfrak{Z} : i

Generalization (2/2)

Voiceless obstruents block agreement in voicing.

$$T = \{ \text{ \mathfrak{Z}, \mathfrak{S}, $\mathfrak{Z},$, \mathfrak{q}} \\ S = \{ \text{ $*\mathfrak{S}\mathfrak{Z}$, $*\mathfrak{S}$, $*\mathfrak{Z}$, $*\mathfrak{Z}$$$

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Voiceless obstruents block agreement in voicing.

Grammar

$$T = \{ \text{ \mathfrak{Z}, \mathfrak{S}, $\mathsf{z}, \mathfrak{J}, $\mathfrak{q}} \}$$

$$S = \{ \text{ *\mathfrak{S}, *\mathfrak{S}, *\mathfrak{S}, *\mathfrak{S}, *\mathfrak{Z}, *$\mathfrak{$$

$$ok = \begin{bmatrix} ok & & & \\ & \ddots & & \\ & & \ddots & \\ & & \ddots & \\ ok & c & & \end{bmatrix}$$

* s a u z:

Generalization (2/2)

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Sibilant Harmony in Implawn Tashlhiyt

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Sibilant Harmony in IMDLAWN TASHLHIYT (Revisited)

ok
 \int q u \mathfrak{Z} : i

Sibilant Harmony in IMDLAWN TASHLHIYT (Revisited)

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$$ok$$
 q 3 :

T₁: sibilant voicing ok q q q q :

Sibilant Harmony in IMDLAWN TASHLHIYT (Revisited)

Sibilant Harmony in IMDLAWN TASHLHIYT (Revisited)

Voiceless obstruents block agreement in voicing:

$$T_1 = \{3, s, z, f, q\}$$
 $S_1 = \{*s_3, *s_2, *g_3, *z_5, *f_2, *f_3, *g_f\}$

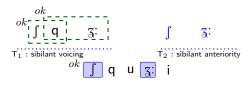
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Sibilant Harmony in IMDLAWN TASHLHIYT (Revisited)

Voiceless obstruents block agreement in voicing:

$$T_1 = \{g, s, z, f, q\}$$
 $S_1 = \{s_3, s_2, s_3, s_4, s_5, s_5, s_7, s_7, s_7\}$

$$T_2 = \{ \mathtt{x}, \ \mathtt{s}, \ \mathtt{z}, \! \} \ S_2 = \{ \mathtt{*s}, \ \mathtt{*s}, \ \mathtt{*g}, \ \mathtt{*g}, \mathtt{*f}, \ \mathtt{*z}, \ \mathtt{*z}, \ \mathtt{*z}, \ \mathtt{*z}, \ \mathtt{*f}, \ \mathtt{*g}, \ \mathtt{*g} \}$$



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Sibilant Harmony in IMDLAWN TASHLHIYT (Revisited)

Voiceless obstruents block agreement in voicing:

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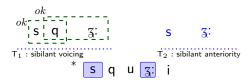
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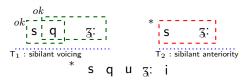


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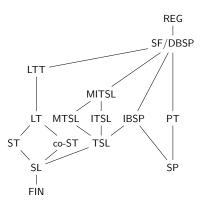


Accounting for Concurrent Processes

► MTSL: TSL closure under intersection (De Santo & Graf, 2019)

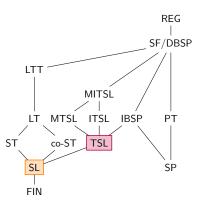
- Intersection closure accounts for multiple concurrent processes
- ► Can characterize the complete phonotactics of a language

A Plethora of Combination



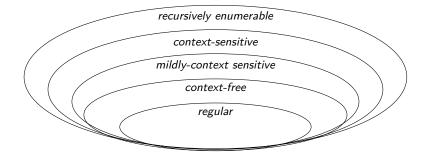
- ► The goal is **not** identifying a single "correct" class
- Pinpoint fundamental properties of the patterns: SL: ▷, TSL: ▷, ...

A Plethora of Combination

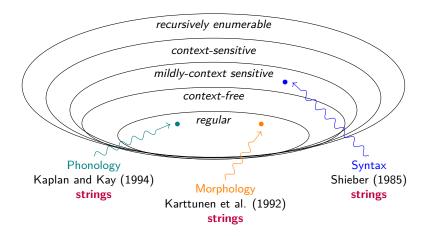


- ► The goal is **not** identifying a single "correct" class
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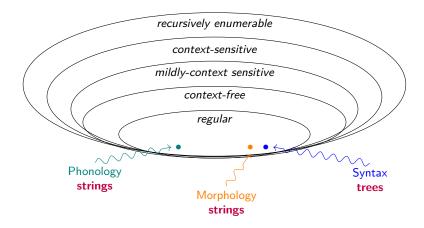
Cross-domain Parallels

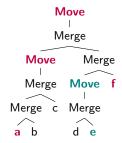


Cross-domain Parallels



Cross-domain Parallels





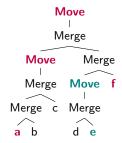
- Some results for syntax
 - regular tree languages
 (Michaelis 2004; Kobele et al. 2007)
 - subregular operations (Graf 2018)
 - subregular dependencies/constraints (Laszakovits 2018; Vu et al. 2019)
 - tree automata and parsing restrictions (Graf & De Santo 19, Ikawa et al. 20)











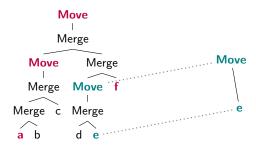
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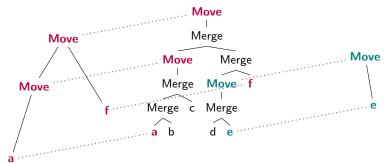
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Interim Summary: Again, So What?

Strong Parallelism

Subregular dependencies in phonology, (morphology), and syntax **subregular** over their respective **structural representations**.

We gain a unified perspective on:

► Attested and unattested typology

learnability?

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 - \times Have a CP iff it dominates ≥ 3 TPs
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Learnable from positive examples of strings/trees. Which information primitives are we sensitive to?

ubregular Classes Cross-Fertilization Artificial Grammar Learning Conclusion

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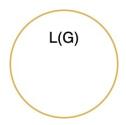
Learnable from positive examples of strings/trees. Which information primitives are we sensitive to?

But:

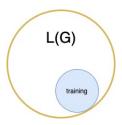
- Typological variation is complex
- Our knowledge of attested pattern is limited

Outline

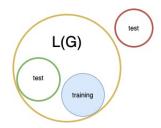
- 1 Linguistics and Formal Language Theory
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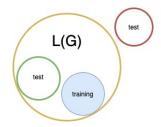
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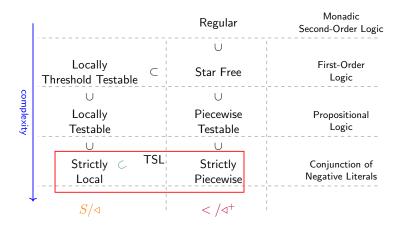


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 → cf. (De Santo and Rawski 2020)

Testing Subregular Predictions



Example: Attested vs. Unattested Patterns

Attested: Unbounded Sibilant Harmony

Every sibilant needs to harmonize

* \$hasxintilaw[\$

ok\$ha[xintilaw[\$

Artificial Grammar Learning

Unattested: First-Last Harmony

Harmony only holds between initial and final segments

ok\$ha**s**xintilaw[\$



* \$satxintilaw[\$

Lai (2015)





Learnable vs. Unlearnable **Harmony Patterns**

Regine Lai

Posted Online July 09, 2015 https://doi.org/10.1162/LING a 00188

@ 2015 Massachusetts Institute of Technology

Linguistic Inquiry Volume 46 | Issue 3 | Summer 2015 p.425-451

Keywords: phonotactics, learnability, computational phonology, formal theory, typology, dependencies

Lai (2015): Stimuli

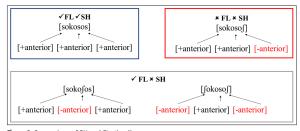


Figure 3: Comparison of SH and FL stimuli.

Lai (2015): Stimuli

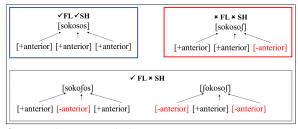


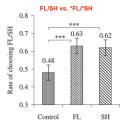
Figure 3: Comparison of SH and FL stimuli.

Table 6

Predicted results with respect to the control group for each test pairing if Sibilant Harmony and First-Last Assimilation grammars were internalized

	Pairs			
	FL/*SH vs. *FL/*SH	FL/SH vs. *FL/*SH	FL/SH vs. FL/*SH	
	(e.g., [s ∫ s] vs.	(e.g., [s s s] vs.	(e.g., [s s s] vs	
Conditions	[s s ∫]) Rate of FL/*SH	[s s ∫]) Rate of FL/SH	$[s \dots \int \dots s]$) Rate of FL/SH	
				SH
FL	> Control	> Control	\sim Control	

Lai (2015): Results



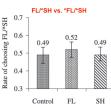


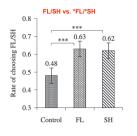
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SH FL	~ Control > Control	> Control > Control	> Control ~ Control	

See Avcu and Hestvik (2020), Avcu et al. (2019) for replications

regular Classes Cross-Fertilization Artificial Grammar Learning Conclusion

Lai (2015): Results



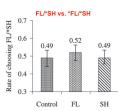


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Conditions	Pairs			
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SH FL	~ Control > Control	> Control > Control	> Control ~ Control	

See Avcu and Hestvik (2020), Avcu et al. (2019) for replications

A Plethora of Testable Predictions

Observation

- ► Attested patterns **A** and **B** are TSL.
- ▶ But combined pattern **A**+**B** is not TSL.

Prediction

► A+B should be harder to learn than A and B

Example: Compounding Markers

Morphotactics as Tier-Based Strictly Local Dependencies

Alëna Aksënova Thomas Graf Sedigheh Moradi

- Russian has an infix -o- that may occur between parts of compounds.
- ► Turkish has a single suffix -sı that occurs at end of compounds.
- (5) vod -o- voz -o- voz water -COMP- carry -COMP- carry 'carrier of water-carriers'
- (6) türk bahçe kapı -sı (*-sı) turkish garden gate -COMP (*-COMP) 'Turkish garden gate'







Example: Compounding Markers [cont.]

Russian and Turkish are TSL.

```
Tier<sub>1</sub> COMP affix and stem edges \# Russian n-grams oo, $0, 0$ Turkish n-grams sisi, $si, si\#
```

- ▶ The combined pattern would yield Ruskish: stem $^{n+1}$ -si n
- ► This pattern is not regular and hence **not TSL either**.

Testable Predictions

Can naive subjects learn Russian-like, Turkis-like, and Ruskish-like compounding?

Outline

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Of Black Swans and Flying Pigs



Of Black Swans and Flying Pigs





Of Black Swans and Flying Pigs





- Not a single data point, but classes of phenomena
- ▶ Value of restrictive theories: predictive and explanatory
- ▶ We learn from falsifying them too!

Complexity as a Magnifying Lens

- ▶ We can compare patterns and predictions across classes
- ▶ We can also compare patterns within a same class

Proceedings of the Society for Computation in Linguistics

Volume 1 Article 8

2018

Formal Restrictions On Multiple Tiers

Alena Aksenova

Stony Brook University, alena.aksenova@stonybrook.edu

Sanket Deshmukh
Stony Brook University, sanket.deshmukh@stonybrook.edu

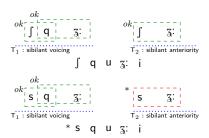




Testing Harmony Systems

Reminder:

► MTSL's multiple-tier idea...



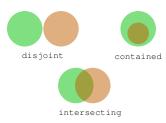


Figure 2: Theoretically possible tier alphabet relations

Testing Harmony Systems (cont.)

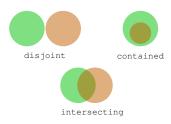
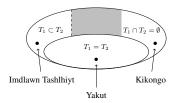


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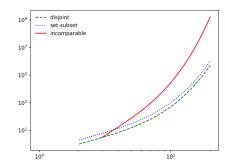


Figure 7: Growth of number of partitions of sets containing up to 20 elements (loglog scale)

Learnability Generalizations

Learning Interactions of Local and Non-Local Phonotactic Constraints from Positive Input

Aniello De Santo Dept. of Linguistics University of Utah aniello.desanto@utah.edu

Alëna Aksënova Google NYC alenaks@google.com

► Efficiently learn MITSL² grammars from positive data

Unlearnable Patterns

- No overlapping tiers with the same ${}^*\rho_1\rho_2$ restriction e.g. $T_1 = \{a, b, c\}, T_2 = \{a, b, d\}, G_1 = G_2 = \{*ab\}$
- ▶ This is *predicted* from the structure of the grammar (see also Lambert et al. 2021)

ubregular Classes Cross-Fertilization Artificial Grammar Learning Conclusion

From Blackbox to Blackbox

Multi-Element Long Distance Dependencies: Using SPk Languages to Explore the Characteristics of Long-Distance Dependencies

Abhijit Mahalunkar

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John D. Kelleher ADAPT Research Center Technological University Dublin Dublin, Ireland john.d.kelleher@dit.ie

- Strictly-piecewise Languages
 - ► Basically: Skip-gram models
 - Capture long distance dependencies over strings
 - Modulate parameters of variation:
 - e.g., length of the dependency, alphabet size, etc.

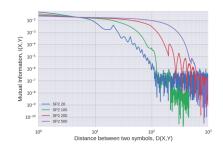


Figure 3: LDD characteristics of datasets of SP2 grammar exhibiting LDDs of length 20, 100, 200 and 500.

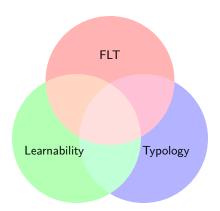
Theory Building

The problem that we cannot deduce [...] theories from data is a limitation, or **perhaps an attribute**, of all empirical science [...] Still, one may abduce hypotheses [...] Abduction is **reasoning from observations** [...] It consists of two steps: generating candidate **hypotheses** (abduction proper), and selecting the "best" explanatory one (inference to the **best explanation**).

(van Roji & Baggio 2020, pg. 9)

Conclusion

A Collaborative Enterprise!



Thank you!



Mathematical Linguistics and Cognitive Complexity

Aniello De Santo, Jonathan Rawski

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References Limits of AGL

From Blackbox to Blackbox

Multi-Element Long Distance Dependencies: Using SPk Languages to Explore the Characteristics of Long-Distance Dependencies

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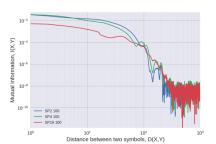


Figure 2: LDD characteristics of datasets of SP2, SP4 and SP16 grammar exhibiting LDD of length 100.

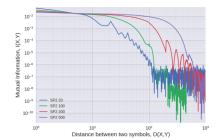


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References Limits of AGL

Example: Circumfixation in Indonesian

Indonesian has circumfixation with no upper bound on the distance between the two parts of the circumfix.

(7) maha siswa big pupil 'student'

- (8) *(ke-) maha siswa *(-an)
 NMN- big pupil -NMN
 'student affairs'
- ▶ Requirements: exactly one ke- and exactly one -an

Tier₁ contains all NMN affixes
Tier₀ contains all morphemes
n-grams \$an, ke\$, keke, anan

References Limits of AGL

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Example: Swahili vyo

b.

Swahili *vyo* is **either a prefix or a suffix**, depending on presence of negation. (?)

```
(9) a. a- vi- soma -vyo
SBJ:CL.1- OBJ:CL.8- read -REL:CL.8

'reads'
```

```
a- si- vyo- vi- soma
SBJ:CL.1- NEG- REL:CL.8- read -OBJ:CL.8
'doesn't read'
```

Example: Swahili vyo [cont.]

```
(10) a. *a- vyo- vi- soma
SBJ:CL.1- REL:CL.8- OBJ:CL.8- read
```

- b. *a- vyo- vi- soma -vyo
 SBJ:CL.1- REL:CL.8- OBJ:CL.8- read -REL:CL.8
- c. *a- si- vyo- vi- soma
 SBJ:CL.1- NEG- REL:CL.8- OBJ:CL.8- read
 -vyo
 REL:CL.8-
- d. *a- si- vi- soma -vyo
 SBJ:CL.1- NEG- OBJ:CL.8- read REL:CL.8-

Example: Swahili vyo [cont.]

Generalizations About vyo

- may occur at most once
- must follow negation prefix si- if present
- ▶ is a prefix iff *si* is present

```
Tier<sub>1</sub> contains vyo, si, and stem edges #
Tier<sub>0</sub> contains all morphemes

n-grams vyovyo, vyo##vyo "at most one vyo"
vyosi, vyo##si "vyo follows si"
si##vyo, $vyo## "vyo is prefix iff si present"
```

TSL Phonology: Accounting for Context

Unbounded Tone Plateauing in Luganda (UTP) No L may occur within an interval spanned by H. (Hyman 2011)

```
(11) a. LHLLLL
b. LLLLHL
c. *LHLLHL
d. LHHHHL
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Input-Sensitive TSL (ITSL) Languages

Defining Tier Projection

Tier projection controlled by:

label of segment

TSL

1

TSL languages are characterized by:

- a 1-local projection function;
- ▶ strictly k-local constraints applied on T.

Input-Sensitive TSL (ITSL) Languages

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1

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- strictly k-local constraints applied on T.

Idea:

- Projection is an input-strictly local transduction Chandlee 2015)
- **What if**: the locality of E_T was higher than 1?

Input-Sensitive TSL (ITSL) Languages

Defining Tier Projection

Tier projection controlled by:

- label of segment
- 2 local context

ITSL
1 + 2
TSL

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Idea:

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Accounting for Context [cont.]

A ITSL analysis for UTP (DeSanto & Graf 2019):

- ▶ Project every H; project L iff immediately follows H
- ▶ Ban: HLH





Accounting for Context [cont.]

- ► Project every H; project L iff immediately follows H
- ► Ban: HLH

```
Example

H

*LHLLHL
```

Accounting for Context [cont.]

- ► Project every H; project L iff immediately follows H
- ► Ban: HLH

```
HL

ok
HL

thicker
```

Accounting for Context [cont.]

- ► Project every H; project L iff immediately follows H
- ► Ban: HLH

Accounting for Context [cont.]

- ► Project every H; project L iff immediately follows H
- ► Ban: HLH

```
HL

ok
LHLLLL

*LHLLHL
```

Accounting for Context [cont.]

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- ► Ban: HLH

```
HL

ok
LHLLLL

*LHLLHL
```

Accounting for Context [cont.]

A ITSL analysis for UTP (DeSanto & Graf 2019):

- ► Project every H; project L iff immediately follows H
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```
|H L|
```

```
*LHLLHL
```

Accounting for Context [cont.]

A ITSL analysis for UTP (DeSanto & Graf 2019):

- ► Project every H; project L iff immediately follows H
- ▶ Ban: HLH

```
H L¦
```

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Accounting for Context [cont.]

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Example

H

Ok L H L L L 
*LH L L H L
```

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Example

H L

*LHL LHL
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HL HL

*LHLLHL
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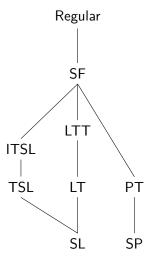
Accounting for Context [cont.]

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Example HL HL ok HL LL * HL HL

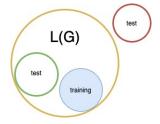
Finer Granularty



Outline

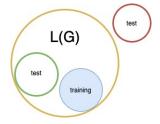
5 Limits of AGL

Testing Predictions with AGL



- It is a powerful technique
- We must be careful in drawing inferences from laboratory behavior
- ▶ Importantly: Common fallacies in experimental design

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The Fallacy of Generalization

Imagine we want to test the ability to learn long-distance dependencies:

Assuming an alphabet $\Sigma = \{a, b, c, d, e\}$, the training samples could look like the following:

```
L_{loc} = \{abcd, aabcd, baacd, bcaae, ...\}

L_{dist} = \{abacd, bacad, bcada, bcaea, ...\}
```

What happens if we test on stimuli with similar distances?

```
L_{test} = \{abcad, abcad, bacda, abcea, \dots\}
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Long-distance relations?



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Limits of AGL

Picking the Right Primitives

Long-distance relations?



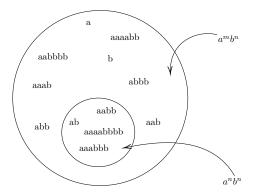
$$\begin{bmatrix} 3 & s \end{bmatrix}$$
 $\begin{bmatrix} a & b & b \\ & 3 & a \end{bmatrix}$ $\begin{bmatrix} a & c & c \\ & 3 & a \end{bmatrix}$ $\begin{bmatrix} b & c \\ & 3 & a \end{bmatrix}$ $\begin{bmatrix} c & c \\ & b \\ & & a \end{bmatrix}$

$$\begin{bmatrix} * \\ \hline 3 \end{bmatrix}$$
a: e r $\begin{bmatrix} \hline s \end{bmatrix}$ e $\begin{bmatrix} ok \\ \hline 3 \end{bmatrix}$ a: e r $\begin{bmatrix} \hline \end{bmatrix}$ $\begin{bmatrix} c \\ \hline \end{bmatrix}$

- Stimuli are often ambiguous between overlapping classes
- Distinguishing between representation requires care

The Set/Subset Problem: Case 1

- ightharpoonup Can participants learn a^nb^n ?
- ightharpoonup We must beware of $a^m b^n$



Evaluating Contrasts

Developmental Constraints on Learning Artificial Grammars with Fixed, Flexible and Free Word Order

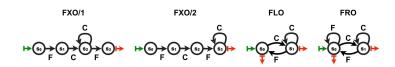
Iga Nowak^{1,2} and 3 Giosuè Baggio^{2,3*}

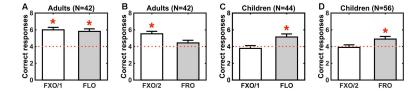
$$FXO/1$$
 $FXO/2$ FLO FRO

$$FXO/2$$
 FLO FRO

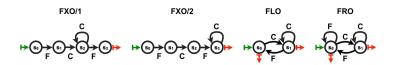
$$FXO/2$$
 FLO FRO

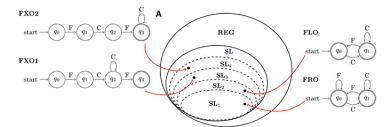
Nowak and Baggio (2017): Results





Complexity Measures and Other Issues (De Santo, 2017)





The Set/Subset Problem: Case 2

Can participants learn a truly free-word order language?

