

General-over-specific markedness bias as a balancing force in GLA-style learning

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The plan!

1. Overview
2. Typology of Balto-Finnic vowel patterns
3. OT analysis of Balto-Finnic vowel patterns
4. Learning, related challenges, and a proposed solution
5. Conclusion

Overview

Grammar-learner relationships

Grammar vs learner, scaled up

Grammar: Constraints needed to account for the patterns of one language

Learner: Settings (parameters, biases) needed to learn a target grammar

... in a vacuum



... situated in a typology



Typology

Balto-Finnic vowel patterns

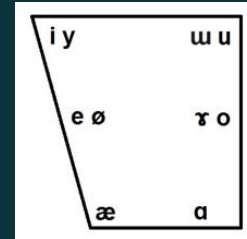
Inventory gaps

Positional restrictions (PR)

Progressive back-front
vowel harmony (VH)

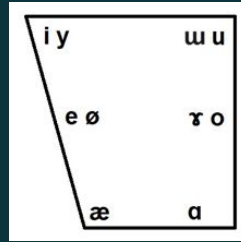
Key references: Kiparsky & Pajusalu (2003),
Bakró-Nagy et al (2022), Vesik (2023).

Balto-Finnic vowel typology



	Language	Inventory	Restrictions outside of $\sigma 1$	Vowel harmony		
				Back	Front	Neutral
PR	N Estonian	i, e, æ, y, ø, ɤ, ɑ, u, o	* æ, y, ø, ʊ, ɤ	-	-	-
	Livonian	i, e, æ, y, ø, ʊ, ɤ, ɑ, u, o	* æ, y, ø, ʊ, ɤ, o	-	-	-
VH	Finnish; Karelian; Ingrian	i, e, æ, y, ø, ɑ, u, o	-	ɑ, u, o	æ, y, ø	i, e
	Votic; Kihnu Est	i, e, æ, y, ø, ɤ, ɑ, u, o	-	ɤ, ɑ, u, o	e, æ, y, ø	i, (e)
PR + VH	N Seto (S Est)	i, e, æ, y, ø, ʊ, ɤ, ɑ, u, o	* ʊ	ʊ, ɤ, ɑ, u, o	e, æ, y, ø	i
	S Seto (S Est)	i, e, æ, y, ø, ʊ, ɤ, ɑ, u, o	* ø, ʊ	ʊ, ɤ, ɑ, u, o	e, æ, y, ø	i, (e), o
	Veps	i, e, æ, y, ø, ɑ, u, o	(* æ, y, ø)	ɑ, u, o	æ, y, ø	i, e

Balto-Finnic vowel typology



Language	Inventory	Restrictions outside of $\sigma 1$	Vowel harmony		
			Back	Front	Neutral
PR N Estonian ←	i, e, æ, y, ø, ʉ, a, u, o	* æ, y, ø, ʉ, a, u, o	-	-	-
Livonian	i, e, æ, y, ø, ʉ, a, u, o	* æ, y, ø, ʉ, a, u, o	-	-	-
Finnish; Karelian; Ingrian	i, e, æ, y, ø, a, u, o	-	a, u, o	æ, y, ø	i, e
Votic; Kihnu Est	i, e, æ, y, ø, ʉ, a, u, o	-	ʉ, a, u, o	e, æ, y, ø	i, (e)
N Seto (S Est)	i, e, æ, y, ø, ʉ, a, u, o	* ʉ	ʉ, a, u, o	e, æ, y, ø	i
S Seto (S Est)	i, e, æ, y, ø, ʉ, a, u, o	* ø, ʉ	ʉ, a, u, o	e, æ, y, ø	i, (e), o
Veps	i, e, æ, y, ø, a, u, o	(* æ, y, ø)	a, u, o	æ, y, ø	i, e

OT Analysis

Balto-Finnic vowel patterns

Stringency + No-disagreement

Constraints - overview

- Context-free markedness constraints for inventory gaps and positional restrictions
- Harmony constraints
- Faithfulness constraints
 - $ID-\sigma_1(\text{Back})$
 >>
 - $ID(\text{Back})$

Stringency relations

Back markedness scale:

- $\omega > \gamma > o > a, u$

Sets & scale-referring constraints:

- $*B_1 = *\{\omega\}$
- $*B_2 = *\{\omega, \gamma\}$
- $*B_3 = *\{\omega, \gamma, o\}$
- $*B_5 = *\{\omega, \gamma, o, a, u\}$

Front markedness scale:

- $\emptyset > \text{æ}, y > e > i$

Sets & scale-referring constraints:

- $*F_1 = *\{\emptyset\}$
- $*F_3 = *\{\emptyset, \text{æ}, y\}$
- $*F_4 = *\{\emptyset, \text{æ}, y, e\}$
- $*F_5 = *\{\emptyset, \text{æ}, y, e, i\}$

No-disagreement

Use local and non-local bans on specific co-occurrences to drive VH. (Pulleyblank, 2002)

- E.g. *BF ; *B...F

No-disagreement

Use local and non-local bans on specific co-occurrences to drive VH. (Pulleyblank, 2002)

- E.g. *BF ; *B...F

Stringency relations + no-disagreement!

E.g.

	*F ₃ <u>B</u> ₅	* <u>B</u> ₅ ...F ₄
a. æ..y..ɯ	*	
b. ɯ..o..æ..e		**

B₅ = {ɯ, ʀ, o, a, u}

F₃ = {∅, æ, y}

F₄ = {∅, æ, y, e}

No-disagreement

Use local and non-local bans on specific co-occurrences to drive VH. (Pulleyblank, 2002)

- E.g. *BF ; *B...F

Stringency relations + no-disagreement!

E.g.

	*F ₃ <u>B</u> ₅	* <u>B</u> ₅ ...F ₄
a. æ..y..ɯ	*	
b. ɯ..o..æ..e		**

B₅ = {ɯ, ʀ, o, a, u}

F₃ = {∅, æ, y}

F₄ = {∅, æ, y, e}

→ Total of 64 harmony constraints, for a total of 64 + 8 + 2 = 74 constraints overall

North Estonian rankings - inventory gaps

- No /w/ in inventory
- $*B_1 \gg \text{ID-}\sigma_1(\text{Bk}) \gg \text{ID}(\text{Bk})$

		$*B_1$	$\text{ID-}\sigma_1(\text{Bk})$	$\text{ID}(\text{Bk})$
/w/				
a.	w	*!		
b.	 i		*	*

Surfaces in e.g. *silm* [silm] “eye”

→ $B_1 = \{w\}$

$B_2 = \{w, r\}$

$B_3 = \{w, r, o\}$

$B_5 = \{w, r, o, a, u\}$

$F_1 = \{\emptyset\}$


$F_3 = \{\emptyset, \text{æ}, y\}$

$F_4 = \{\emptyset, \text{æ}, y, e\}$

$F_5 = \{\emptyset, \text{æ}, y, e, i\}$

North Estonian rankings - positional restrictions

- No /æ, y, ø, ʊ, ʏ/ in non-initial syllables
- $ID-\sigma_1(Bk) \gg *F_3$; $*B_2 \gg ID(Bk)$

		$ID-\sigma_1(Bk)$	$*F_3$	$ID(Bk)$
/æ..y/				
a.	æ..y		**!	
b.	 æ..u		*	*
c.	a..u	*!		**

Surfaces in e.g. *käru* [kæru] “cart”

$B_1 = \{\omega\}$

→ $B_2 = \{\omega, \gamma\}$

$B_3 = \{\omega, \gamma, o\}$

$B_5 = \{\omega, \gamma, o, a, u\}$

$F_1 = \{\emptyset\}$

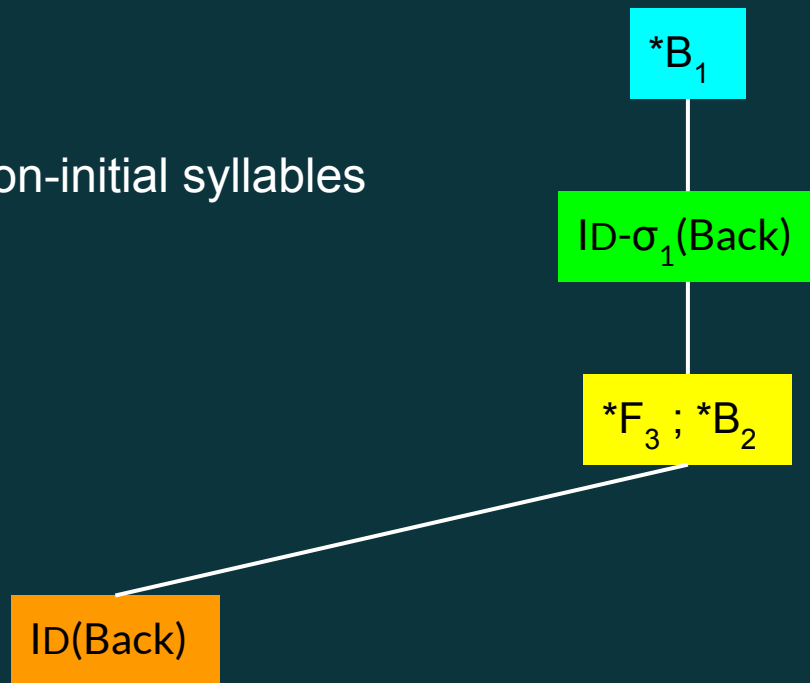
→ $F_3 = \{\emptyset, \text{æ}, y\}$

$F_4 = \{\emptyset, \text{æ}, y, e\}$

$F_5 = \{\emptyset, \text{æ}, y, e, i\}$

North Estonian rankings – overall

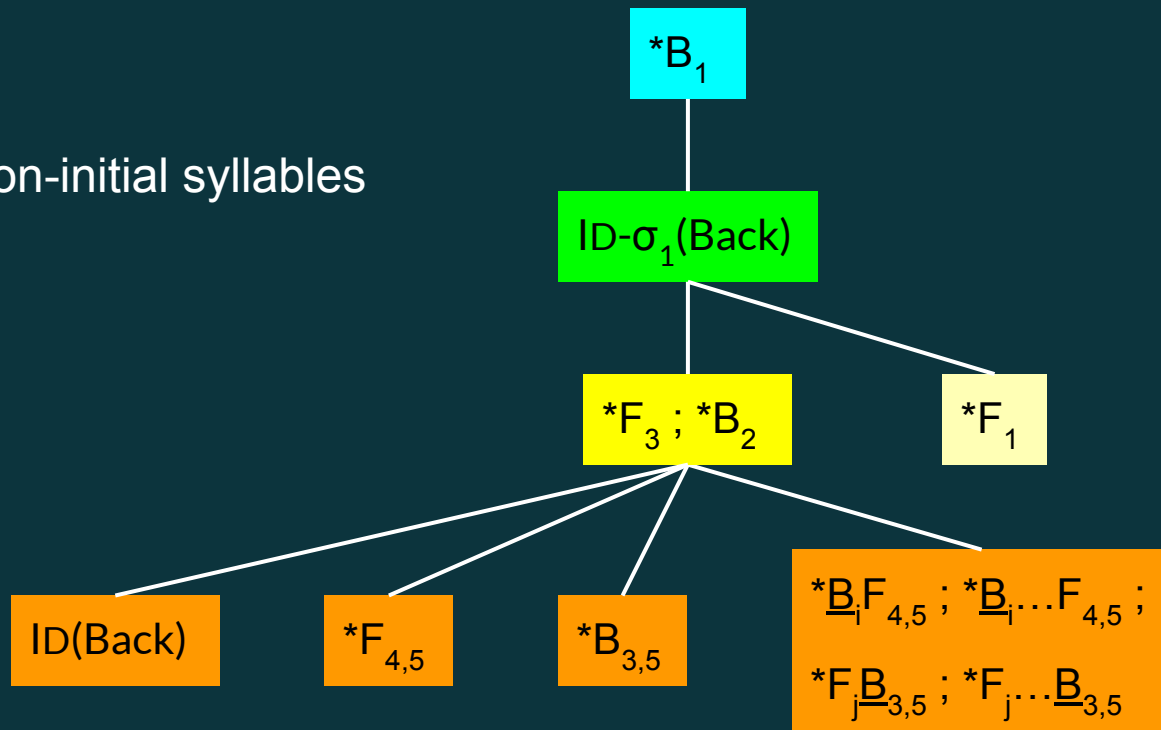
- No /w/ in inventory
- No /æ, y, ø, ω, ʏ/ in non-initial syllables



North Estonian rankings – overall

- No /w/ in inventory
- No /æ, y, ø, ω, ɣ/ in non-initial syllables

*B_iF_{1,3} ; *B_i...F_{1,3}
 ;
 *FB_{j,2} ; *F_j...B_{1,2}



Learning

Simulations

Obstacles

A novel proposal



Learning is hard

Biases widely
used

+

Biases for this
typology

Even with all the usual suspects for biases, a GLA¹-type learner can't learn this simple North Estonian grammar from positive evidence.

- How does it fail?
- What can we do?

¹ OT-based Gradual Learning Algorithm (Boersma & Hayes, 2001)

Learning – basic setup

Commonly-used settings:

- Low initial faithfulness (Smolensky, 1996)
M constraints start at 100 / F at 0
- Specific over general faithfulness (Hayes, 2004)
 $ID-\sigma_1(\text{Back}) \geq ID(\text{Back}) + 20$

Learning - basic setup

Learning data:



“õlu”
[ɣlu]



- Simulated North Estonian
- Positive evidence only
(i.e. identity-mapped inputs)

/ɣlu/	*B ₁	*B ₂	*F ₃	⋮	Id-σ ₁ (Bk)	Id(Bk)
a. ɣlu		*		⋮		
b. ɣly		*	*	⋮		*
c. elu				⋮	*	*
d. ely			*	⋮	*	**

Learning - basic setup

Bias needed to address other learning challenges in this typology:

- Update rule with tempered promotion rate
- $\textit{promotion rate} = \frac{1}{1+W} \times \textit{demotion rate}$ (Magri & Kager, 2015)
- $W = \#$ of winner-preferring constraints

Learner A

As proposed:

- Constraint set
- Commonly-used settings/biases
- Tempered promotion rate

Learner A - Results (excerpts)

Ideal results:

*B₁

>>

ID-σ₁(B_k)

>>

*F₃; *B₂

>>

ID(B_k); *F_{4,5}; *B_{3,5}; *B_iF_{4,5}; *B_i...F_{4,5}; *F_jB_{3,5}; *F_j...B_{3,5}

Actual results:

constraint	final ranking value
*B1	100.000
*F5B2	100.000
*F5...B2	100.000
*B5F3	100.000
*B5...F3	100.000
Id-σ1(B _k)	83.021
*F1	65.286
*B2	63.267
Id(B _k)	63.021
*F3	23.457
*B3	3.859
*F4	-40.324
*F5	-40.324
*B5	-40.656
*B5...F5	-47.456
*B5F5	-49.727
*F5B5	-84.667
*F5...B5	-102.667

Learner A - Results (excerpts)



Ideal results:

*B₁

>>

Id-σ₁(Bk)

>>

*F₃; *B₂

>>

Id(Bk); *F_{4,5}; *B_{3,5}; *B_iF_{4,5}; *B_i...F_{4,5}; *F_jB_{3,5}; *F_j...B_{3,5}


Actual results:

- *F₃ below Id(Bk)
- Coincidental VH
- Success rate on test evaluations: 0.7865



constraint	final ranking value
*B1	100.000
F5B2	100.000
*F5...B2	100.000
*B5F3	100.000
*B5...F3	100.000
Id-σ ₁ (Bk)	83.021
*F1	65.286
*B2	63.267
Id(Bk)	63.021
*F3	23.457
*B3	3.859
*F4	-40.324
*F5	-40.324
*B5	-40.656
*B5...F5	-47.456
*B5F5	-49.727
*F5B5	-84.667
*F5...B5	-102.667

Learner A - Results

Under the grammar acquired by Learner A...

	$*\underline{\mathbf{B}}_5 \dots \mathbf{F}_3$	$\text{ID-}\sigma_1(\text{Bk})$	$\text{ID}(\text{Bk})$	$*\mathbf{F}_3$
/u..ɑ/				
a.  u..ɑ				
b. u..æ	*!		*	*

A grammatical UR surfaces faithfully.

	$*\underline{\mathbf{B}}_5 \dots \mathbf{F}_3$	$\text{ID-}\sigma_1(\text{Bk})$	$\text{ID}(\text{Bk})$	$*\mathbf{F}_3$
/y..æ/				
a.   y..æ				**
b. y..ɑ			*!	*

An ungrammatical UR also surfaces faithfully.

A novel proposal

Idea: Give general M constraints a chance to get credit for the phonotactics of the target grammar.

- Ensures maximal restrictiveness.

Inspiration: Albright & Hayes (2006) determine initial ranking based on generality.

- Prevents super-specific “junk” constraints (induced by the MGL) from taking too much credit.
- Allows more general constraints to be active where possible.

Implementation: Initial articulated hierarchy of markedness constraint values.

- Function of each constraint’s rate of application in a sample set of inputs.
- Can be freely reversed by learning data.

$M_{\text{gen}} \gg M_{\text{spec}}$ bias

During pre-learning “observation” stage:

- Learning rate = 0
- Learner is fed randomly-sampled inputs
- Learner keeps a running total of the number of times each markedness constraint is violated by the heard inputs

$M_{\text{gen}} \gg M_{\text{spec}}$ bias

After observation stage and before learning:

- Calculate average generality for each M

$$g_M = \frac{\text{\# violations of M}}{\text{total \# inputs}}$$

- Calculate initial ranking value for each M

Initial $\theta_M = 100(b + mg_M)$

where g_M = generality for constraint M
 b = y-intercept coefficient
 m = slope coefficient

Learner B

Same as Learner A but with this change:

- Markedness constraints distributed by generality

$$\theta_M = 100(b + mg_M) \quad b = 1.0, m = 1.0$$

Learner B - Initial hierarchy

Same as Learner A but with this change:

- Markedness constraints distributed by generality

$$\theta_M = 100(b + mg_M) \quad b = 1.0, m = 1.0$$

constraint	initial ranking value
*B5	254.280
*F5	228.860
*F4	182.420
*F5..._B5	170.700
*B3	159.520
*F5_B5	153.640
*_B5...F5	146.660
*_B5F5	137.780
*F3	133.900
*F1	111.800
*B2	111.320
*B1	100.000
*F5_B2	100.000
*F5..._B2	100.000
*_B5F3	100.000
*_B5...F3	100.000

Learner B - Results (excerpts)

Ideal results:

*B₁

>>

ID-σ₁(B_k)

>>

*F₃; *B₂

>>

ID(B_k); *F_{4,5}; *B_{3,5}; *B_iF_{4,5}; *B_i...F_{4,5}; *F_jB_{3,5}; *F_j...B_{3,5}

Actual results:

constraint	final ranking value
*B1	110.667
*F5B2	100.000
*F5...B2	100.000
*B5F3	100.000
*B5...F3	100.000
Id-σ1(B _k)	96.941
*B2	88.243
*F1	86.771
*F3	80.563
Id(B _k)	76.941
*F5	68.419
*F4	49.199
*B3	48.106
*B5	45.902
*B5...F5	-32.551
*B5F5	-39.249
*F5...B5	-114.758
*F5B5	-127.038

Learner B - Results (excerpts)

Ideal results:

*B₁

>>

ID-σ₁(B_k)

>>

*F₃; *B₂

>>

ID(B_k); *F_{4,5}; *B_{3,5}; *B_iF_{4,5}; *B_i...F_{4,5}; *F_jB_{3,5}; *F_j...B_{3,5}


Actual results:

- Align with ideal!
- Success rate on test evaluations: 0.9837


constraint	final ranking value
*B1	110.667
*F5B2	100.000
*F5...B2	100.000
*B5F3	100.000
*B5...F3	100.000
Id-σ1(B _k)	96.941
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*B5F5	-39.249
*F5...B5	-114.758
*F5B5	-127.038

Learner B - Results

Under the grammar acquired by Learner B...

	$*\underline{B}_5 \dots F_3$	ID- σ_1 (Bk)	$*F_3$	ID(Bk)
/u..ɑ/				
a.  u..ɑ				
b. u..æ	*!		*	*

A grammatical UR surfaces faithfully.

	$*\underline{B}_5 \dots F_3$	ID- σ_1 (Bk)	$*F_3$	ID(Bk)
/y..æ/				
a. y..æ			**!	
b.  y..ɑ			*	*

$*F_3 \gg ID-\sigma_1$ (Bk) ensures that an ungrammatical UR is repaired.

Conclusion

Grammar-learner relationships

Summary

Takeaways



Summary

Main ideas from today:

- Constraints with competing violation profiles cause challenges with respect to the Credit Problem (Dresher, 1999).
- *General-over-specific markedness* bias as counterpart to *specific-over-general faithfulness* helps to offset potential adverse effects
- Specifically, it allows a GLA-type learner to avoid falling into the subset problem resulting from learning a less-restrictive grammar (Jesney & Tessier, 2011)

Summary

This talk zooms in on just a mere slice of my dissertation.



Summary

This talk zooms in on just a mere slice of my ~~dissertation~~.
dessert-ation?



Summary

This talk zooms in on just a mere slice of my dissertation. The whole cake involves:

- The entire Balto-Finnic vowel pattern typology
- The parameters and biases that are needed to make sure that any language's patterns can be learned in that broader context
- The general-markedness bias being crucial to the most successful learners



Takeaways

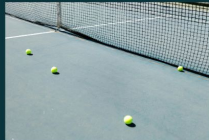
Takeaways

Grammar vs learner, scaled up

Grammar: Constraints needed to account for the patterns of one language

Learner: Settings (parameters, biases) needed to learn a target grammar

... in a vacuum



<https://www.pexels.com/photo/tennis-ball-on-tennis-court-8224638/>

... situated in a typology



<https://www.picknik.com/tennis-exercise-playground-ball-sport-hobby-134795>

Takeaways

Specific-faithfulness bias is often used to maintain restrictiveness

- General-markedness bias can help too

Takeaways

Specific-faithfulness bias is often used to maintain restrictiveness

- General-markedness bias can help too

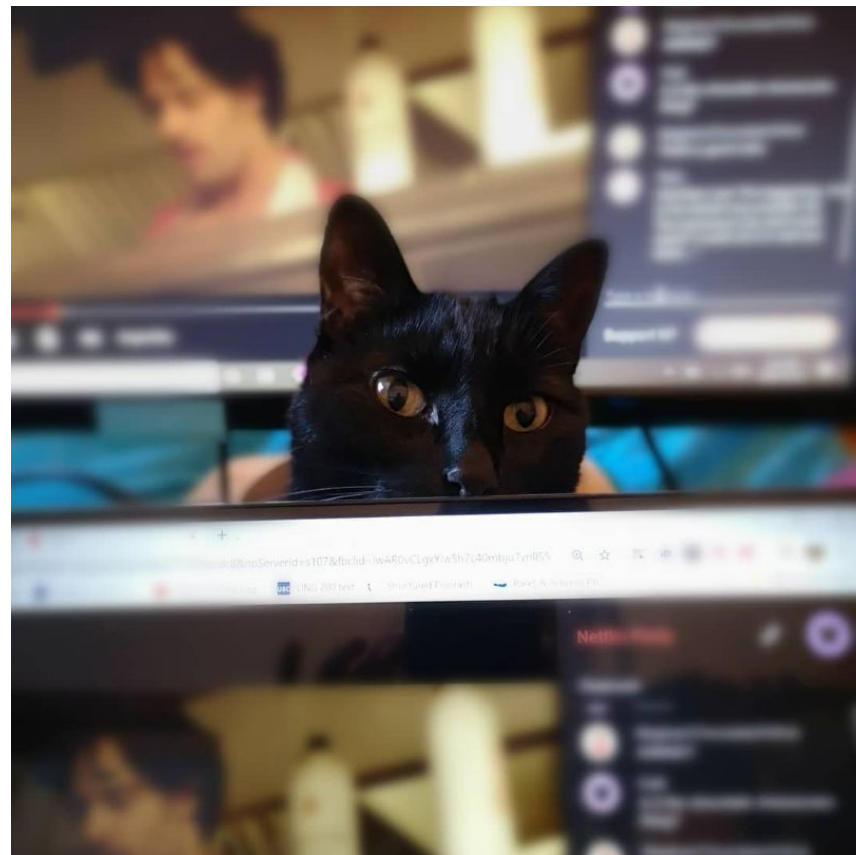
Related problems could arise in other contexts in which constraints are similar enough to compete for credit/blame

- By definition
- By accident / sheer numbers (what does reality look like)?



Thank you!

- To all of you 😊
- UBC Phonology Discussion Group and Child Phonology Lab members
- AMP 2022 & 2023 audiences



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Typological examples

Inventory gaps

E.g. Finnish: *{u, ɣ}

- varpunen [ɑ..u..e] (*sparrow*) *[ɑ..u..ɣ]

Positional restrictions

E.g. N Estonian: *{æ, y, ø, ɣ}

- küla [y..ɑ] (*village*) *[y..æ]

Back-front vowel harmony

E.g. North Seto: transparent {i}

- lusikka [u..i..ɑ] (*spoon*) *[u..i..æ]

Typological tangents 1

From [Slide 7: Balto-Finnic vowel typology](#) re variably transparent vowels

South Seto

- /e/ as first-syllable vowel does not trigger front harmony
- Non-initial /e/ is found only in front-harmonic words

Kihnu Estonian

- /e/ sometimes triggers front harmony & sometimes doesn't
- Can vary even within a single lexical item e.g. mõttes~mõttõs [ɤ..e]~[ɤ..ɤ] (*in thought*)

Lindstrom (2013)

	Language(s)	Inventory	Restrictions in σ_{2+}	Vowel harmony			
				Back	Front	Transp't	Opaque
(d)	Votic; Kihnu Est	i, e, æ, y, ø, ɤ, a, u, o	-	ɤ, a, u, o	e, æ, y, ø	i, (e)	-
(f)	S Seto (S Est)	i, e, æ, y, ø, ʉ, ɤ, a, u, o	* ø, ʉ	ʉ, ɤ, a, u, o	e, æ, y, ø	i, (e)	o

Typological tangents 2

From [Slide 7: Balto-Finnic vowel typology](#) re opaque vowels

South Seto

- /o/ as first-syllable vowel triggers back harmony
- Non-initial /o/ interrupts front harmony and begins a back-harmonic domain

E.g. hämonõ [æ..o..ɤ] (*bleary*) *[æ..o..e] Pajusalu (2022: 369)

	Language(s)	Inventory	Restrictions in σ_{2+}	Vowel harmony			
				Back	Front	Transp't	Opaque
(f)	S Seto (S Est)	i, e, æ, y, ø, ʉ, ɑ, u, o	* ø, ʉ	ʉ, ɤ, ɑ, u, o	e, æ, y, ø	i, (e)	o

Typological tangents 3

From [Slide 7: Balto-Finnic vowel typology](#) re decaying vowel harmony

Veps

- VH no longer productive past first few syllables (precise domain is unclear)
- Vowels in later syllables restricted to /i, e, a, u, o/
- E.g. tütär [y..æ] (*girl*) Grünthal (2015: 45)
külmetumaa [y..e..u..a] (*we get cold*) Grünthal (2015: 47)

	Language(s)	Inventory	Restrictions in σ_{2+}	Vowel harmony			
				Back	Front	Transp't	Opaque
(g)	Veps	i, e, æ, y, ø, a, u, o	(* æ, y, ø)	a, u, o	æ, y, ø	i, e	-

“Missing” stringency sets

From [Slide 11: Stringency relations](#)

- Go back to slide

Opacity as a function of focus

From [Slide 14: No-disagreement](#)

- Focus on B ensures that front harmony goes as far as possible (no “sour grapes”)

	/æ..u..o..y/	Id-σ1(Bk)	*B5F4	*B5...F4	*F1	*F3B5	*F3...B5	Id(Bk)
a) Front harmony all the waaaaaay!	æ..y..ø..y				*			**
b) Front harmony until /o/, then back	æ..y..o..u 🙌					*	**	**
c) Sour grapes (why bother changing the /u/ if we're going to switch to back anyway?)	æ..u..o..u					*	***!	*

MaxEnt learner (UCLA-PL)

From [Slide 20: Learning is hard](#)

- Could also consider a MaxEnt learner (e.g. UCLA Phonotactic Learner; Hayes & Wilson, 2008)
- No URs and therefore no faith constraints at all; just a probability distribution over possible surface forms
- Induces feature-based co-occurrence constraints, given:
 - Inputs consisting of surface forms only
 - Feature values of each segment
- Can't seem to figure out opacity because it doesn't have a way to induce "focus"

3 learners on 3 sample languages

Final grammars' success rates on test evaluations:

	N Estonian (PR)	Finnish (VH)	N Seto (PR + VH)
Learner \emptyset	0.2455	0.2683	0.2984
Learner A	0.7865	1.0000	0.9988
Learner B	0.9837	1.0000	0.9999