A weak but flexible logic for modeling phonological processes

Adam Jardine
Rutgers University

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Collaborators

Siddharth Bhaskar (U. of Southern Denmark)

Jane Chandlee (Haverford)

Tatevik Yolyan (Rutgers)

Overview

Overview

Boolean monadic recursive schemes (BMRS) [1]

- Restricted fragment of recursive programs[2]
- Capture the subsequential functions[3] on strings
- Useful for describing phonological processes[4]
- Current work focusing on learning[5]

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[1](Bhaskar et al. 2020)
[2](Moschovakis 2019)
[3](Schützenberger 1977; Mohri 1997)
[4](Chandlee and Jardine 2021)
[5](Yolyan 2025)

What is phonology?

Phonology as functions

English plural

"-z"		"-s"	
[piz]	‘peas’	[kʰlæsps]	‘clasps’
[tʰoʊz]	‘toes’	[mɪts]	‘mitts’
[dɑlz]	‘dolls’	[bloʊks]	‘blokes’
[dɑgz]	‘dogs’	[kʰɑfs]	‘coughs’
[læbz]	‘labs’		

Phonology as functions

English plural

"-z"		**"-s"**	
[piz]	'peas'	[kʰlæsps]	'clasps'
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[dɑlz]	'dolls'	[bloʊks]	'blokes'
[dɑgz]	'dogs'	[kʰɑfs]	'coughs'
[læbz]	'labs'		

Analysis:

- **Underlying representation (UR)** of plural: /-z/
- A **process** mapping URs to **surface representations (SRs)**: (/piz/,[piz]), (/dɑlz/,[dɑlz]), (/kʰlæspz/,[kʰlæsps]), (/mɪtz/,[mɪts]), …

Goals of Generative phonology

$$(/\text{piz}/,[\text{piz}]),\ (/\text{dɑlz}/,[\text{dɑlz}]),\ (/\text{k}^{\text{h}}\text{læspz}/,[\text{k}^{\text{h}}\text{læsps}]),$$
$$(/\text{mɪtz}/,[\text{mɪts}]),\ \ldots$$

- How do we capture phonological processes?
  - SPE[6]: [-sonorant] $\rightarrow$ [-voice] / [-voice] __
  - OT[7]: Agree in voicing $\gg$ Don't change voicing
- What is a *possible* phonological process?

[6] Sound Pattern of English (Chomsky and Halle 1968)
[7] Optimality Theory (Prince and Smolensky 1993)

Goals of Generative phonology

High-tone spreading in Shambaa (Odden 1982)

[ku-ʃunt^{h}-a]	‘to wash’	(/σσσ/, [σσσ])
[ku-ɣoʃo-a]	‘to do’	(/σσσσ/, [σσσσ])
[ku-tʃí-ʃúnt^{h}-a]	‘to wash it’	(/σσ́σσ/, [σσ́σ́σ])
[ku-ví-ɣóʃó-a]	‘to do them’	(/σσ́σσσ/, [σσ́σ́σ́σ])
[ku-ɣoʃo-a-ɣoʃo-a]	‘to do repeatedly’	(/σσσσσσσσ/, [σσσσσσσσ])
[ku-tʃí-ɣóʃó-á-ɣóʃó-a]	‘to do it repeatedly’	(/σσ́σσσσσσσ/, [σσ́σ́σ́σ́σ́σ́σσ])

Goals of Generative phonology

Long-distance consonant harmony in Kikongo

[sos-ila]	‘searched for’	(/sos-ila/,	[sos-ila])
[sakid-ila]	‘congratulate for’	(/sakid-ila/,	[sakid-ila])
[ku-to:t-ila]	‘to harvest for’	(/ku-to:t-ila/,	[ku-to:t-ila])
[ku-kin-ina]	‘to dance for’	(/ku-kin-ila/,	[ku-kin-ina])
[ku-dumuk-ina]	‘to jump for’	(/ku-dumuk-ila/,	[ku-dumuk-ina])
[ku-dumuk-is-ina]	‘to make jump for’	(/ku-dumuk-is-ila/,	[ku-dumuk-is-ina])

Goals of Generative phonology

- Why?
  - ✓ High tone spreads to second-to-last syllable
  - × High tone spreads to center syllable

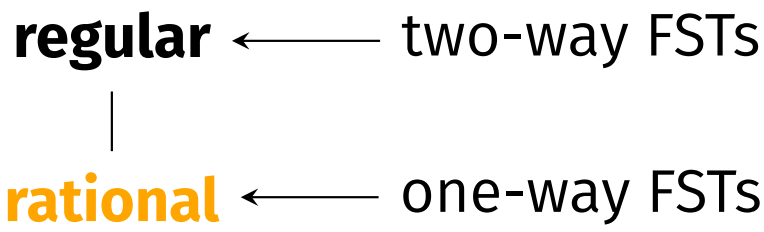
Phonology and computation

What kinds of functions are phonological processes?

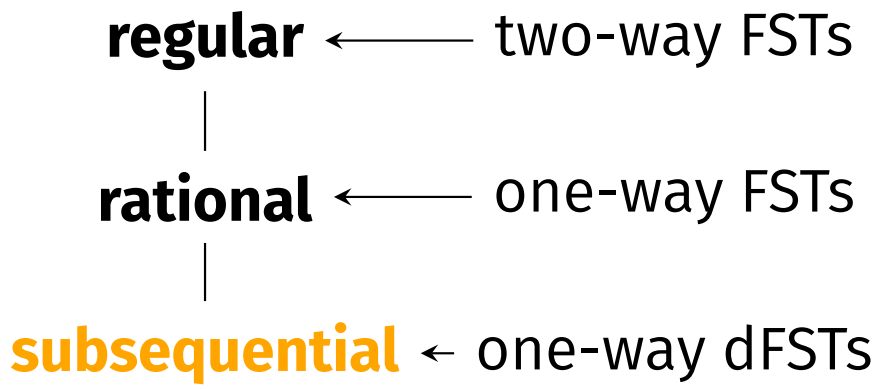
- Heinz (2007);Heinz (2018): Phonological generalizations have a *computationally* restrictive character

What kinds of functions are phonological processes?

- Johnson (1972);Kaplan and Kay (1994):
  Phonological rules are **rational**



What kinds of functions are phonological processes?



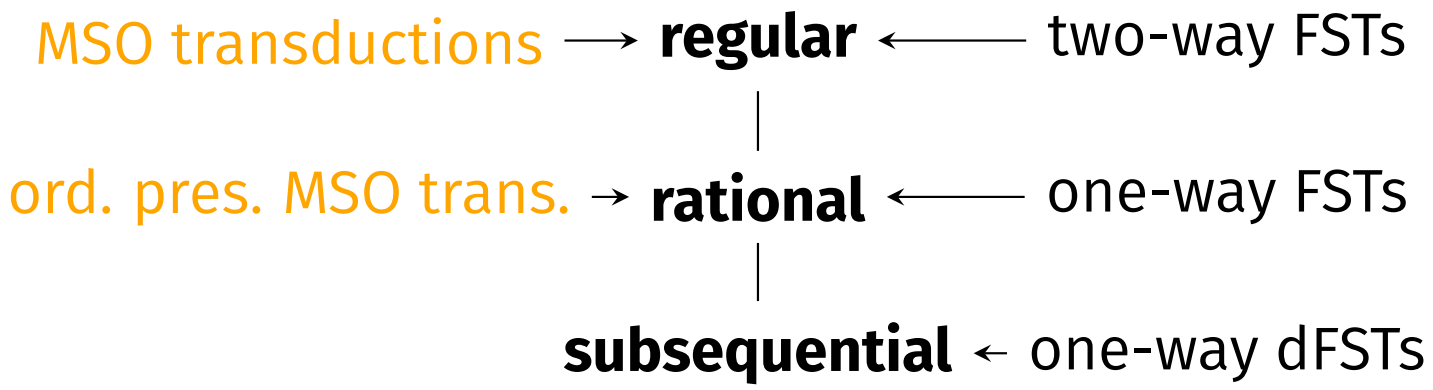
- Mohri (1997);Heinz and Lai (2013):
  Phonological processes are **subsequential**

What kinds of functions are phonological processes?

- Heinz (2018): Connections between logic and FLT[8] are useful for phonology
  - Ex: features, syllable structure, etc.

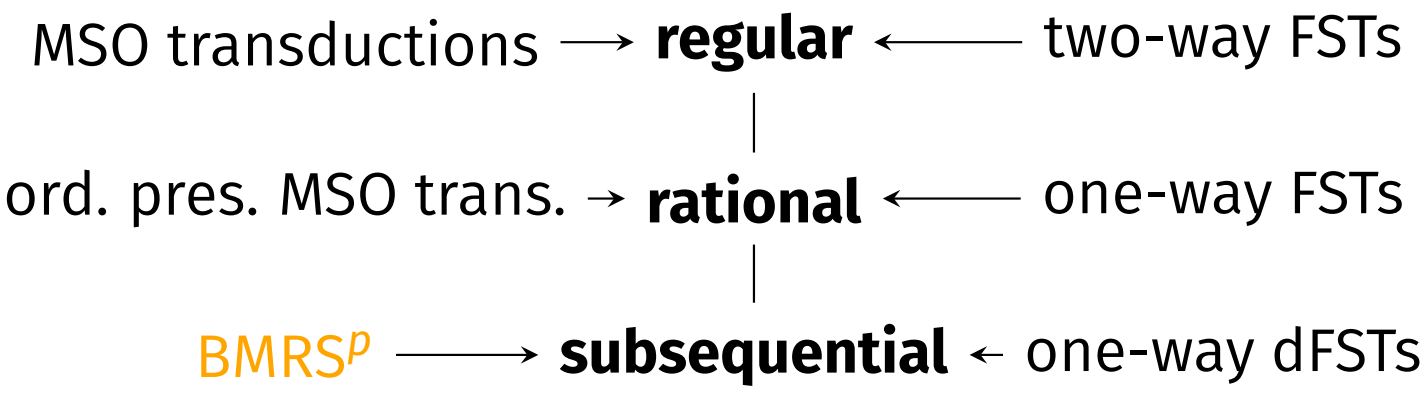
[8] Büchi (1960), McNaughton and Papert (1971), et. seq.

What kinds of functions are phonological processes?



Filiot and Reynier (2016)

What kinds of functions are phonological processes?



Bhaskar et al. (2020)

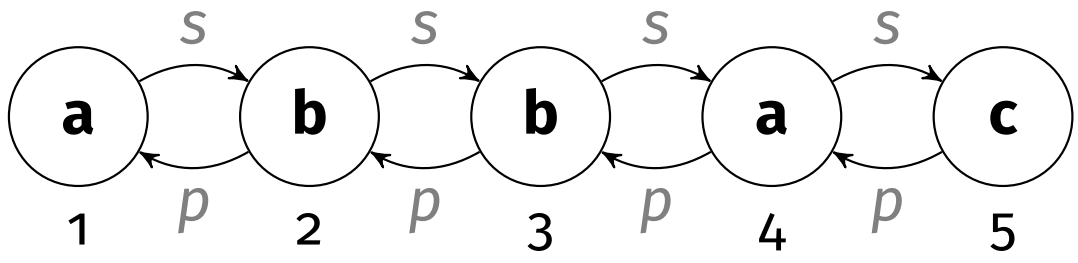
Boolean Monadic Recursive Schemes

BMRS

- Associate strings in Σ^* with structures of the form

$$\langle D; P_\sigma \text{ for } \sigma \in \Sigma, p, s \rangle$$

- E.g., *abbac* would be



where $P_a = \{1, 4\}$, $P_b = \{2, 3\}$, $P_c = \{c\}$.

BMRS

- $\mathtt{X}$ are index variables; $\mathtt{F}$ are function variables
- Syntax:

$$T \to \mathtt{x} \mid \top \mid \bot \mid \mathtt{f}(T) \mid s(T) \mid p(T) \mid \sigma(T) \mid \mathtt{if}\ T\ \mathtt{then}\ T\ \mathtt{else}\ T$$

  where:

  - $\mathtt{x} \in \mathtt{X}$, $\mathtt{f} \in \mathtt{F}$, $\sigma \in \Sigma$
  - $\mathtt{x}$, $s(T)$ and $p(T)$ are index-typed, all others boolean
  - For $\sigma(T)$ and $\mathtt{f}(T)$, T must be index-typed
  - For $\mathtt{if}\ T\ \mathtt{then}\ T\ \mathtt{else}\ T$, each T is boolean-typed

BMRS

- Syntax:

$$T \to \quad \mathtt{x} \mid \top \mid \bot \mid \mathtt{f}(T) \mid s(T) \mid p(T) \mid \\ \sigma(T) \mid \mathtt{if}\ T\ \mathtt{then}\ T\ \mathtt{else}\ T$$

- Boolean predicates with one free variable, e.g.

$$\mathtt{if}\ a(\mathtt{x})\ \mathtt{then}\ b(p(\mathtt{x}))\ \mathtt{else}\ \bot$$

- Assume $\mathtt{f}(T)$ and $\sigma(T)$ return $\bot$ if T is undefined (at end or beginning of word)

BMRS

- A recursive, boolean, monadic **scheme** (or *program*) pairs $\vec{\mathtt{f}} = (\mathtt{f}_1, \ldots, \mathtt{f}_n)$ with definitions:

$$\begin{aligned} \mathtt{f}_1(\mathtt{x}) &= T_1(\vec{\mathtt{f}}, \mathtt{x}) \\ &\vdots \\ \mathtt{f}_n(\mathtt{x}) &= T_n(\vec{\mathtt{f}}, \mathtt{x}) \end{aligned}$$

  where each T_i is boolean.

- The semantics is given by least-fixed semantics of recursive programs

BMRS

An example

- Let $\Sigma = \{a, b, c\}$ and let $\vec{f} = (\mathtt{a}, \mathtt{b}, \mathtt{c})$.

$$\mathtt{a}(\mathtt{x}) = \mathtt{if}\ \mathtt{b}(\mathtt{x})\ \mathtt{then}\ \bot\ \mathtt{else}\ a(\mathtt{x})$$

$$\mathtt{b}(\mathtt{x}) = \mathtt{if}\ b(\mathtt{x})\ \mathtt{then}\ \top\ \mathtt{else}\ \mathtt{if}\ a(\mathtt{x})\ \mathtt{then}\ \mathtt{b}(p(\mathtt{x}))\ \mathtt{else}\ \bot$$

$$\mathtt{c}(\mathtt{x}) = c(\mathtt{x})$$

BMRS

An example

$$\mathtt{a}(\mathtt{x}) = \mathtt{if}\ \mathtt{b}(\mathtt{x})\ \mathtt{then}\ \bot\ \mathtt{else}\ a(\mathtt{x})$$

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$$\mathtt{c}(\mathtt{x}) = c(\mathtt{x})$$

input: $a \quad b \quad a \quad a \quad c \quad a$

BMRS

An example

$$\mathtt{a}(\mathtt{x}) = \mathtt{if}\ \mathtt{b}(\mathtt{x})\ \mathtt{then}\ \bot\ \mathtt{else}\ a(\mathtt{x})$$

$$\mathtt{b}(\mathtt{x}) = \mathtt{if}\ b(\mathtt{x})\ \mathtt{then}\ \top\ \mathtt{else}\ \mathtt{if}\ a(\mathtt{x})\ \mathtt{then}\ \mathtt{b}(p(\mathtt{x}))\ \mathtt{else}\ \bot$$

$$\mathtt{c}(\mathtt{x}) = c(\mathtt{x})$$

input: a b a a c a

$\mathtt{a}(\mathtt{x})$

$\mathtt{b}(\mathtt{x})$

$\mathtt{c}(\mathtt{x})$

BMRS

An example

$$\mathtt{a}(\mathtt{x}) = \mathtt{if}\ \mathtt{b}(\mathtt{x})\ \mathtt{then}\ \bot\ \mathtt{else}\ a(\mathtt{x})$$

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$$\mathtt{c}(\mathtt{x}) = c(\mathtt{x})$$

input:	a	b	a	a	c	a
a(x)	$\top$					
b(x)	$\bot$					
c(x)	$\bot$					

BMRS

An example

$$\mathtt{a}(\mathtt{x}) = \mathtt{if}\ \mathtt{b}(\mathtt{x})\ \mathtt{then}\ \bot\ \mathtt{else}\ a(\mathtt{x})$$

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input:	a	b	a	a	c	a
a(x)	$\top$	$\bot$				
b(x)	$\bot$	$\top$				
c(x)	$\bot$	$\bot$				

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input:	a	b	a	a	c	a
a(x)	$\top$	$\bot$				
b(x)	$\bot$	$\top$	$\top$			
c(x)	$\bot$	$\bot$				

BMRS

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input:	a	b	a	a	c	a
$\mathtt{a}(\mathtt{x})$	$\top$	$\bot$	$\bot$			
$\mathtt{b}(\mathtt{x})$	$\bot$	$\top$	$\top$			
$\mathtt{c}(\mathtt{x})$	$\bot$	$\bot$	$\bot$			

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$$\mathtt{c}(\mathtt{x}) = c(\mathtt{x})$$

input:	a	b	a	a	c	a
$\mathtt{a}(\mathtt{x})$	$\top$	$\bot$	$\bot$	$\bot$		
$\mathtt{b}(\mathtt{x})$	$\bot$	$\top$	$\top$	$\top$		
$\mathtt{c}(\mathtt{x})$	$\bot$	$\bot$	$\bot$	$\bot$		

BMRS

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$$\mathtt{c}(\mathtt{x}) = c(\mathtt{x})$$

input:	a	b	a	a	c	a
a(x)	$\top$	$\bot$	$\bot$	$\bot$	$\bot$	$\top$
b(x)	$\bot$	$\top$	$\top$	$\top$	$\bot$	$\bot$
c(x)	$\bot$	$\bot$	$\bot$	$\bot$	$\top$	$\bot$

BMRS

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$$\mathtt{c}(\mathtt{x}) = c(\mathtt{x})$$

input:	a	b	a	a	c	a
$\mathtt{a}(\mathtt{x})$	$\top$	$\bot$	$\bot$	$\bot$	$\bot$	$\top$
$\mathtt{b}(\mathtt{x})$	$\bot$	$\top$	$\top$	$\top$	$\bot$	$\bot$
$\mathtt{c}(\mathtt{x})$	$\bot$	$\bot$	$\bot$	$\bot$	$\top$	$\bot$
output:	a	b	b	b	c	a

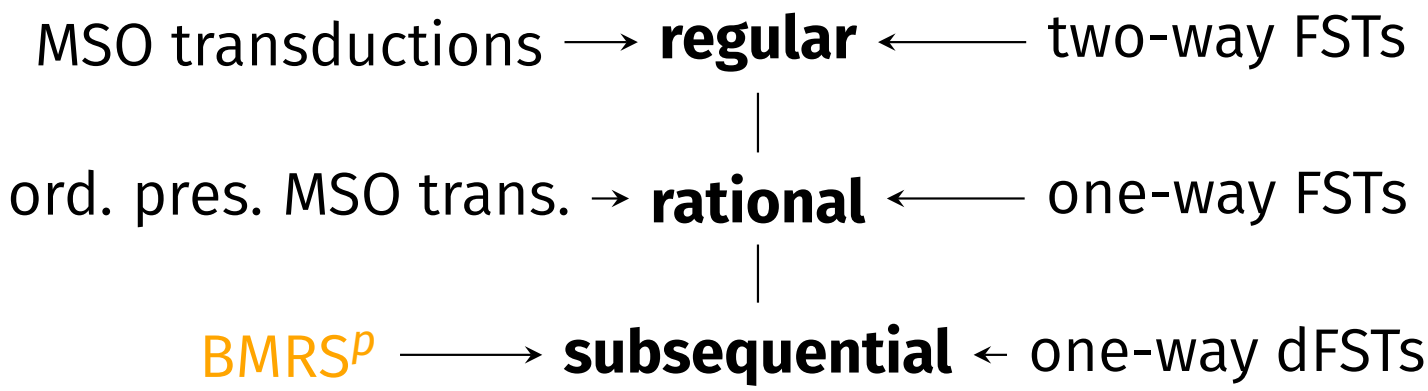
BMRS

- Fix an **output alphabet** Γ. We can include a function symbol in $\vec{\mathtt{f}}$ for each $\gamma \in \Gamma$.

$$\begin{gathered}
\gamma_1(\mathtt{x}) = T_1(\vec{\mathtt{f}}, \mathtt{x}) \\
\vdots \\
\gamma_k(\mathtt{x}) = T_k(\vec{\mathtt{f}}, \mathtt{x}) \\
\mathtt{f}_1(\mathtt{x}) = T_{k+1}(\vec{\mathtt{f}}, \mathtt{x}) \\
\vdots \\
\mathtt{f}_n(\mathtt{x}) = T_{k+n}(\vec{\mathtt{f}}, \mathtt{x})
\end{gathered}$$

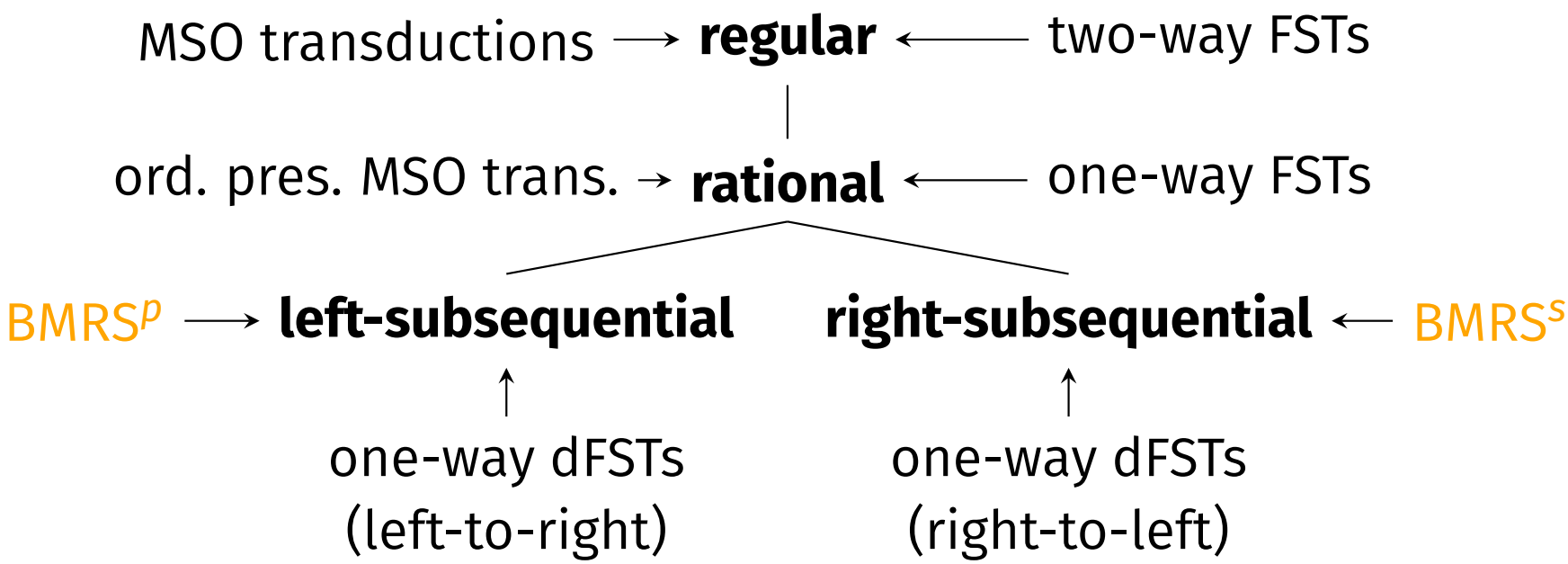
- We can interpret this BMRS as a transduction from $\Sigma^* \to \Gamma^*$

BMRS - Expressivity



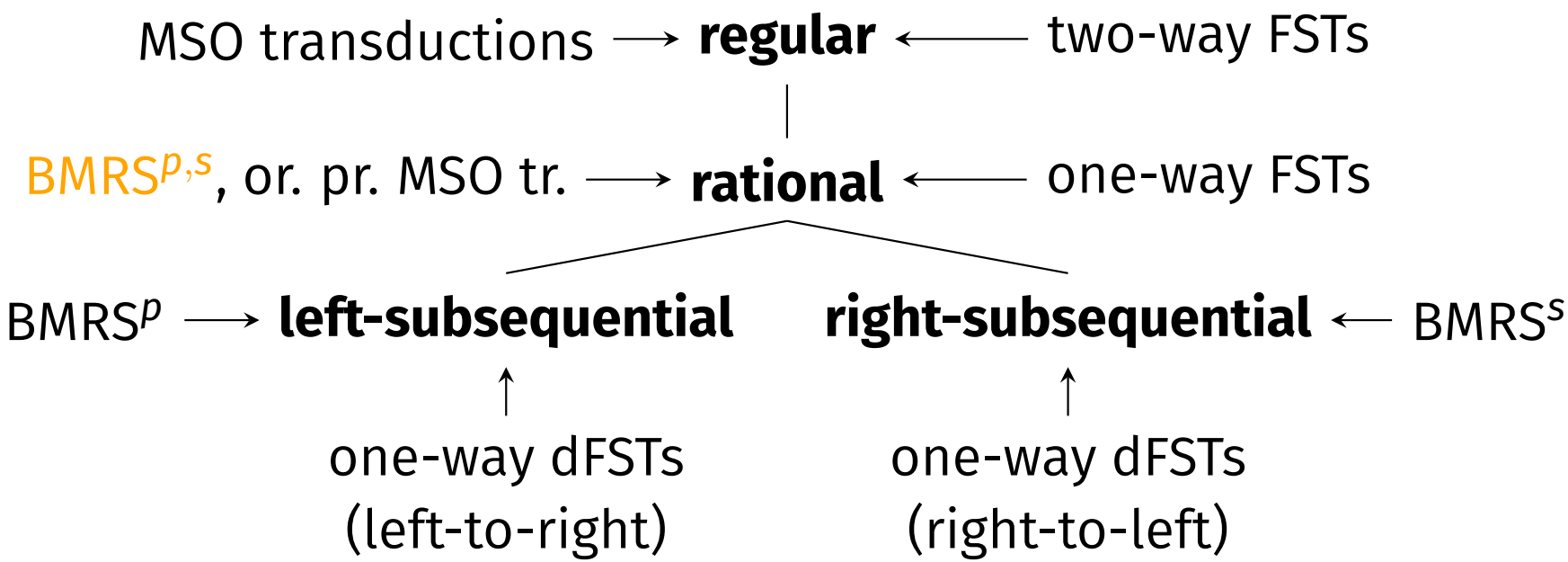
Bhaskar et al. (2020)

BMRS - Expressivity



Bhaskar et al. (2020)

BMRS - Expressivity



Bhaskar and Jardine (in prep.)

BMRS and phonology

Shambaa spreading[9]

$$/\sigma\sigma\sigma\sigma\sigma/ \mapsto [\sigma\sigma\sigma\sigma\sigma]$$
$$/\sigma\acute{\sigma}\sigma\sigma\sigma/ \mapsto [\sigma\acute{\sigma}\acute{\sigma}\acute{\sigma}\sigma]$$

$$\acute{\sigma}'(\mathtt{x}) = \mathtt{if}\ \acute{\sigma}(\mathtt{x})\ \mathtt{then}\ \top\ \mathtt{else}$$
$$\mathtt{if}\ \mathtt{last}(\mathtt{x})\ \mathtt{then}\ \bot\ \mathtt{else}\ \acute{\sigma}'(p(\mathtt{x}))$$
$$\sigma'(\mathtt{x}) = \mathtt{if}\ \acute{\sigma}'(\mathtt{x})\ \mathtt{then}\ \bot\ \mathtt{else}\ \top$$

[9] We can include $\mathtt{last}(T)$ to input signature without increasing expressivity

BMRS and phonology

$$\acute{\sigma}'(\mathrm{x}) = \texttt{if } \acute{\sigma}(\mathrm{x}) \texttt{ then } \top \texttt{ else}$$
$$\texttt{if last}(\mathrm{x}) \texttt{ then } \bot \texttt{ else } \acute{\sigma}'(p(\mathrm{x}))$$
$$\sigma'(\mathrm{x}) = \texttt{if } \acute{\sigma}'(\mathrm{x}) \texttt{ then } \bot \texttt{ else } \top$$

- Captures 'Elsewhere Condition" effects (Chandlee and Jardine 2021; Jardine and Oakden 2025)
- Even long-distance phonological processes are 'myopic' (Wilson 2003, 2006; though c.f. McCollum et al. 2020)

Learning

Learning

- dFSTs are effectively learnable (Oncina, García, and Vidal 1993)
- Yolyan (2025): fragments of BMRS^p are learnable

$$\begin{array}{cc|c} a & b & a \\ & & \downarrow \\ a & b & b \end{array}$$

To conclude

To conclude

- BMRS are a flexible, yet well-behaved model for computing string transductions
  - BMRS$^{p} \rightarrow$ left-subsequential
  - BMRS$^{s} \rightarrow$ right-subsequential
  - BMRS$^{p,s} \rightarrow$ rational
- They are useful for describing phonology
- There is a lot to learn about learning BMRS
- Also BMRS and B-RASP?

Thank you!

References I

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