# Modeling morphological productivity with the Minimal Generalization Learner

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# Outline

#### Model overview

- Minimal generalization
- Reliability
- Confidence

#### A simple simulation

- Phonological rules
- 'Impugnment'
- Non-local and suprasegmental contexts

Minimal generalization Reliability Confidence

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Image: A matched black

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# Goal of the model

• Given: a set of morphologically related forms—e.g., English:

Present	Past		Present	Past	
mıs	mıst	'miss(ed)'	nid	nidəd	'need(ed)'
prɛs	prɛst	'press(ed)'	dʒ∧mp	dʒ∧mpt	ʻjump(ed)'
læf	læft	'laugh(ed)'	plæn	plænd	'plan(ned)'
h∧g	h∧gd	'hug(ged)'	sıŋ	sæŋ	'sing/sang'
глb	rʌbd	'rub(bed)'	drıŋk	dræŋk	'drink/drank'

- Goal: generalize to new items, such as novel verb [splink]<sub>Present</sub>
  - Expected outputs

Very likely/acceptable: splrŋkt Somewhat likely/acceptable: splʌŋk, splæŋk Quite unlikely/unacceptable: splɔt

- Not:
  - \*splinkd, \*splinkad (misapplication of phonology)
  - \*sploon (valid past, wrong context)
  - \*splind (idiosyncratic change: make~made)

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# **Desired** information

#### Some questions

- What string mappings relate the morphological categories? (Ø →d, I→æ, Iŋk→ɔt, etc.)
- Are some mappings phonological variants of others? ( $\emptyset \to d, \emptyset \to t, \emptyset \to ad$ )
- Are some mappings restricted (categorically, probabilistically) to specific phonological contexts?
- What is the relative strength of the mappings in various contexts?

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### Induction strategy

- Parse pairs to discover changes and contexts—e.g.,
  - mis  $\sim$  mist = arnothing 
    ightarrow t / mis \_\_\_\_ #
  - dʒʌmp  $\sim$  dʒʌmpt = arnothing 
    ightarrow t / dʒʌmp \_\_\_\_ #
- Compare contexts to discover broader contexts
  - mis, dʒʌmp: /[-voi]\_\_\_\_
- Evaluate accuracy of resulting contexts

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### **Discovering changes**

Pinker & Prince (1988), p. 130:

- Candidates for rules are hypothesized by comparing base and past tense versions of a word, and factoring apart the changing phonetic portion, which serves as the rule operation, from certain morphologically-relevant phonological components of the stem, which serve to define the class of stems over which the operation can apply." (= the context)
  - Pinker and Prince (1988), Ling and Marinov (1993), Albright and Hayes (2002)

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# A simple parsing strategy

Albright and Hayes (2002, 2003)

- Edge-in alignment: start from both left and right, aligning until there's a mismatch
- Combine leftward and rightward alignments to align as many segments as possible

 Medial changes: leftward and rightward scans are both able to align some segments; residue left in the middle

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# Factoring out changes and contexts

- Mismatched portions = change (A  $\rightarrow$  B)
- Aligned portions = context (C \_\_\_\_ D)



Image: A matrix and a matrix

• Resulting rule: arnothing o t / #m1s \_\_\_\_ #

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# Word-specific rules

Result: a set of word-specific rules

a. 
$$\emptyset \rightarrow t$$
 / #mis \_\_\_ #  
b.  $\emptyset \rightarrow t$  / #pres \_\_\_ #  
c.  $\emptyset \rightarrow t$  / #læf \_\_\_ #  
d.  $\emptyset \rightarrow d$  / #hAg \_\_\_ #

e. 
$$\emptyset \rightarrow d$$
 / #rʌb \_\_\_ #

f. 
$$\varnothing \rightarrow ad / #nid _ #$$

g. 
$$\emptyset \rightarrow t / #d_3 mp \___ #$$

h. 
$$\emptyset \rightarrow d$$
 / #plæn \_\_\_\_ #

i. 
$$I \rightarrow a e / s \_ n #$$
  
j.  $I \rightarrow a e / dr \_ n k #$ 

Adam Albright (albright@mit.edu) MGL Tutorial (11/64)

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# Comparing rules

- miss and press both take  $\varnothing \to t$ . Assume that this is because they have some crucial property in common
  - In both cases, the change is word-final
  - In both cases, the segment before the change is an [s]
  - In both cases, the segment before [s] is a non-low lax front V
  - etc...??? (sonorant before the vowel, monosyllabic, etc.)
- Albright and Hayes (2002): Minimal generalization approach
  - Conservative collapsing: new rule keeps *everything* that the pair has in common

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# Minimal generalization

• Once again, some alignment scheme is needed. Assume locality

	m	I	s	 #
р	r	ε	s	 #

Precludes many possible generalizations, such as:

- End in /s/
- Vowel is front, lax
- First segment is labial
- A pragmatic assumption: myopic generalization
  - Once a mismatch is encountered and featural generalization is needed, shared similarities farther away from the change location are not likely to be crucial
  - Convert more distant segments to free variables (X, Y)
  - We'll come back to this assumption later

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Image: A matched black

# Minimal generalization

#### Comparing miss and press: $\varnothing ightarrow$ t / ...

	Residue	Shared	Shared	Change	Shared	Shared	Residue
		Features	Segments	Location	Segments	Features	
Α.	#m	I	S		#		
В.	#pr	ε	S		#		
C.	х	+syllabic -low -back -tense -round	S		#		

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# Minimal generalization

Minimal generalization of features: retain all shared feature values

- Prevents generalization to unseen feature values
- Permits generalization to unseen feature combinations
- Example: comparing b and n



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### Examples

#### Comparing hug, plan, and then rub: $\varnothing ightarrow \mathsf{d}$ / ...

	Residue	Shared	Shared	Change	Shared	Shared	Residue
		Features	Segments	Location	Segments	Features	
Α.	#h∧	g			#		
В.	#plæ	n			#		
C.	Х	-syllabic       -continuant       -labial       -lateral       +voice			#		

#### And generalizing further with rub:

D.	#r∧	b	#
E.	х	-syllabic       -continuant       -lateral       +voice	#

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# Examples

#### Comparing sing and drink: $I \rightarrow a m / ...$

	Residue	Shared	Shared	Change	Shared	Shared	Residue
		Features	Segments	Location	Segments	Features	
Α.	#	S			ŋ		#
В.	#d	r			ŋ		k#
C.	Х	-syllabic +coronal +continuant			ŋ		Y

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Image: A matrix and a matrix

#### What this model does well

Rapidly discovers most general environment for each pattern



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Minimal generalization Reliability Confidence

Image: Image:

# Rapid generalization

- Generalization proceeds rapidly, given sufficiently diverse stems
- The pathways shown here can be found using verbs that are among the 75 most frequent verbs of English (according to CELEX) (of which the majority are actually irregular)

Minimal generalization Reliability Confidence

### Rule creation for irregulars

"Irregular" mappings are also compared and generalized



 Provides rules to generalize [splɪŋ]→[splæŋ] in addition to [splɪŋd]

Minimal generalization Reliability Confidence

Image: A matched black

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### What determines the strength of a rule?

- Goal: suffixed output [splɪŋd] is more probable/acceptable than outputs like [splæŋ], [splʌŋ], etc.
- Irregular patterns (at least in English) tend to cover relatively few forms, which are similar to one another  $\approx$  share a set of phonological properties
- Result: rules that characterize them are more specific, and weaker strength
- These patterns are not productive, except possibly for inputs that fit a very particular phonological shape ( $\approx$  prototypical forms)

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# What determines the strength of a rule?

Pinker and Prince (1988):

• "Rule candidates increase in strength each time they have been exemplified by an input pair"

	Rule	Examples covered
1.	$\varnothing ightarrow$ t / #mis #	1
2.	arnothing ightarrow t / #prɛs #	1
3.	arnothing  ightarrow t / X [+syl, $-$ low, $-$ bk, $-$ tns, $-$ rnd] s #	2
4.	arnothing  — d / #hʌg #	1
5.	arnothing ightarrow d / #plæn #	1
6.	$\varnothing  ightarrow$ d / X [-syl,-cont,-lab,-lat,-del.rel.]	2
7.	$ ext{i}  ightarrow  extbf{æ}$ / #s ŋ#	1
8.	$ extsf{i}  ightarrow  extsf{æ}$ / #dr ŋk#	1
9.	$ ext{i}  ightarrow$ æ / X [-syl, +cor, +cont] ŋ Y#	2

• p. 134: "[by various means], some regularities can be enshrined as permanent productive rules whereas others can be discarded or treated differently."

A B > A B >

Minimal generalization Reliability Confidence

# Simply counting examples is not enough

• Consider hypothesized rules after exposure to sing, ring, drink, sit



- "Prototypical" forms (*sing*, *ring*, *drink*) and "outlier" (*sit*) combine to yield more general rule
- More general rule encompasses more forms, so has higher strength

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Simply counting examples is not enough

- Prediction:  ${\rm I}{\rightarrow} \varpi$  change may apply equally well to any form within this broader context
  - Hypothetical zick  $\sim$  zack just as well supported as spling  $\sim$  splang

Image: A mathematical states and a mathem

Minimal generalization Reliability Confidence

Image: A matrix and a matrix

# Diagnosis

- *Strength*, measured purely in terms of number of examples covered, favors the broadest possible generalizations
- Under this view, "compatible with the data" = "encompasses all the examples"
- Broad generalizations are not always virtuous; sometimes, they incur numerous exceptions.
- A more intuitive notion of "compatible with the data" must incorporate not only generality, but also accuracy

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### Rule reliability

Reliability of a hypothesized rule:

number of items that the rule works for ("hits")

number of items that meet the rule's structural description (CAD) ("scope")

#### Also known as accuracy, or coverage



- Island of reliability: context in which a change is especially likely
  - "especially likely" = more likely than in its most general context (Albright 2002)

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# The intended effect

- Favoring rules with high reliability should let the model find specific contexts that *uniquely* characterize members of a particular class
- Overly broad contexts are discouraged, because of the cost of adding exceptions
  - Number of positive hits gained (numerator) must exceed number of exceptions added to scope (denominator)
  - For small, "irregular" class, optimal contexts will be narrow
  - For "regular" classes, there may be fewer distinguishing features that uniquely characterize most regular verbs, so no benefit to staying small

- Model zooms to full generality, incurring some exceptions
- Model should settle on the right level of generality for each class

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### How this plays out in practice

- Example: ran the model on the 200 most frequent verbs in CELEX
  - 134 regular (80 -*d*, 30 -*t*, 24 -*əd*); 66 irregular, of various types

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# How this plays out in practice

- Example: ran the model on the 200 most frequent verbs in CELEX
  - 134 regular (80 -*d*, 30 -*t*, 24 -*əd*); 66 irregular, of various types
- Most reliable rules for a few of the major classes

Rule	Hits	Scope	Rel.	Examples	Exceptions	Excludes
i: $\rightarrow \epsilon / \# \begin{bmatrix} -syl \\ +cont \end{bmatrix} \_ d#$	3	3	1.00	read, lead, feed	(none)	meet
- conditidants						
Rule	Hits	Scope	Rel.	Examples	Exceptions	Excludes
$e \rightarrow o / \# \begin{bmatrix} -syl \\ +voi \\ +lab \end{bmatrix} \_ r$	2	2	1.00	wear, bear	(none)	break
= b, m, v, r						
Rule	Hits	Scope	Rel.	Examples	Exceptions	Excludes
	2	2	1.00	set, let	(none)	put, cut, get, sit, 
Rule	Hits	Scope	Rel.	Examples	Exceptions	Excludes
	2	3	0.67	send, spend	tend	build
$= p,t,k,f,\theta,s,f,tf$						
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# How this plays out in practice

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Image: Image:

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- Example: ran the model on the 200 most frequent verbs in CELEX
  - 134 regular (80 -*d*, 30 -*t*, 24 -*əd*); 66 irregular, of various types
- Most reliable rules for a few of the major classes

Rule	Hits	Scope	Rel.	Examples	Exceptions	Excludes
arnothing ightarrow d / X ə' #	11	11	1.00	remember, consider, offer,	(none)	69 other [+voi]
$\varnothing \to d / X \begin{bmatrix} +syl \\ -low \\ +back \end{bmatrix} v \_ #$	5	5	1.00	 move, love, serve, prove, remove	(none)	75 other [+voi]
$egin{aligned} \end{aligned} \end{aligned}$	5	5	1.00	remain, explain, plan, join, contain	(none)	75 other [+voi]

Minimal generalization Reliability Confidence

Image: A matrix and a matrix

#### The fate of the more general rules

#### The more general rules are not nearly so reliable

Rule	Hits	Scope	Rel.	Examples	Exceptions
$\varnothing  ightarrow$ d / X [+voi] #	80	134	.60	seem, use, try, call, turn	do, say, go, know, see, <b>need</b>
$\varnothing \rightarrow t$ / X [-voi] #	30	66	.46	look, ask, work, talk, help,	get, think, take, <b>want</b> , put,
$ \overrightarrow{\varnothing} \rightarrow \operatorname{ad} / X \begin{bmatrix} -\operatorname{syl} \\ +\operatorname{cons} \\ -\operatorname{son} \\ -\operatorname{cont} \\ -\operatorname{strid} \\ +\operatorname{cor} \\ +\operatorname{ant} \\ = \{t, d\} \end{bmatrix} $	24	45	.53	want, need, start, wait, expect,	get, find, put, sit, stand, 

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### An unintended consequence

- Focusing on reliability has something of the desired effect
  - Finds consists clusters of similar words
  - Identifies cases where it more distant words should be excluded
- However, it takes small-scale generalizations too seriously
- By favoring reliability instead of generality, we end up finding a long list of very accurate, but seemingly quite accidental contexts
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### Why are the more general contexts so unreliable?

- Genuine exceptions: almost half of *t*, *d*-final verbs in this sample are irregular
  - Expect this proportion to go down in a larger sample
  - Larger sample will never make the problem disappear completely, however, since some subcontexts are 100% regular
    - E.g., verbs ending in voiceless fricatives, verbs ending in [as], etc.
    - These will always be more reliable than the general context
  - In any system with irregularity, broad generalizations will necessarily involve some exceptions (decreased reliability)
- Inadequate characterization of rule interaction:
  - t, d-final verbs act as exceptions for more general -d, -t suffixes

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More on this below...

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# Confidence

- Reliability is defined as a proportion
  - 2/2 = 7/7 = 100/100
- Speaker intuitions: not all 'perfectly reliable' rules are equally productive
- 100/100 less likely to be coincidental than 2/2

Minimal generalization Reliability Confidence

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### A comparison

Reliability values from a larger set of English verbs:



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# Confidence

Implementation: lower confidence intervals

- Although 4/4 = 1, still likely that the next example will be an exception
  - Confidence interval: actual proportion might be somewhat lower or higher than observed
  - In this case, we can be 95% sure that the true value is between .875 and 1.125 (if values above 1 made sense here)
- Lower confidence limits:



Minimal generalization Reliability Confidence

### The effects of confidence

- Patterns involving small numbers of words may be weak, even if consistent
- Morphological categories that are rare may yield overall less confident rules

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Image: A matched black

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# The training data

#### Text file of training pairs and test items

- Training pairs (and optionally, token frequency)
- Test forms for 'wug testing'
- Phonological features
  - .fea file with feature specifications
  - "ASCII" column (numeric codes): values don't matter, should just be unique

- Each segment must be a single character (ASCII, or Unicode)
- Feature values: 0-1 (binary), 0-n (scalar)
- Unspecified = -1

### Running the model

- Java archive: MinGenLearner.jar
  - From command line: java -jar MinGenLearner.jar
- Output files (tab-delimited text format)
  - .out file with list of changes
  - .sum file with wug test results
  - .rules file with list of rules discovered (if 'Save rules' option selected)

Image: A matrix and a matrix

Phonological rules 'Impugnment' Non-local and suprasegmental contexts

Image: A matched black

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## Insufficient generalization without phonology

Most general 'regular' rules in the sample simulation



- Misses fact that t, ad could be derived by phonological rules
  - d ightarrow t / [-voi] \_\_\_\_

• 
$$arnothing
ightarrow$$
 ə / {t,d} \_\_\_\_ d

Phonological rules 'Impugnment' Non-local and suprasegmental contexts

Image: A matrix and a matrix

# Unifying changes with phonology

#### Goal:

- Allow model to discover that adding *d* after other contexts would yield illegal sequences such as \*pd, \*dd
- Posit phonological rules that repair illegal sequences to generate the attested output

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# Approach to discovering phonological rules

- Provide model with list of phonotactically (surface) illegal sequences
  - \*pd, \*kd, \*sd, \*td, \*dd, \*bt, \*gt, \*zt, \*tt, \*dt, ...
  - Listed in .ill file
- Doppelgänger' rules
  - $\bullet~$  When positing a rule A  $\to$  B / C \_\_\_ D, try out other changes A'  $\to$  B' in the same context C \_\_ D
  - Original:  $\varnothing \to t$  / [-voi] \_\_\_\_ #
  - Doppelgänger arnothing
    ightarrow d / [-voi] \_\_\_\_ #
- Scan outputs of Doppelgänger rules for illegal sequences
- If difference between illegal output and attested output is a possible phonological mapping, posit a rule
  - $\bullet~$  Change/insertion/deletion of a single segment—e.g., d  ${\rightarrow}t$
- Model options: 'Use Phonology', 'Use Doppelgängers'

A B > A B >

# Some limitations

- Can't discover rule ordering
- Can't detect conflicting rules to see which is the productive one
  - kd  $\rightarrow$  kt: l $\upsilon$ kd  $\rightarrow$  l $\upsilon$ kt 'look')
  - kd  $\rightarrow$  d: meikd  $\rightarrow$  meid 'make'
- Assumes that rules hold in entire string (no NDEB or morphological conditioning)
  - pd  $\rightarrow$  pt: <code>vpdeitad</code>  $\rightarrow$  \*<code>vpteitad</code> 'update'
  - Workaround: add word boundaries and augment list of illegal sequences to restrict to final position (\*pd#, \*dt#, etc.)
- Also possible to pre-specify mappings by brute force in . phon file
  - End-run around limited discovery mechanism

Phonological rules 'Impugnment' Non-local and suprasegmental contexts

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### Results for sample file

Most general 'regular' rules, with phonology enabled



- Model discovers more general  $\varnothing 
  ightarrow$ d rule
- Reliability of most general  $\varnothing 
  ightarrow {
  m t}$  rule improves, too

Phonological rules 'Impugnment' Non-local and suprasegmental contexts

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## Another problem

#### Model outputs for [splɪŋ]

-	Output	Rule	Rel.	Conf.		
-	splɪŋd	$\varnothing \to d / X \begin{bmatrix} -syl \\ 1-2son \\ 0-1aper \end{bmatrix} $	50/65	.729		
= f,m,n,s,v,z,δ,ŋ,∫,ʒ,θ						
	splɪŋt		79/142	.528		
	splæŋ	$ ext{i}  ightarrow  extbf{æ}$ / X ' {m,n,ŋ}	2/4	.310		
• $\varnothing \rightarrow t$ rule covers voiceless segments $+$ n (I $\Rightarrow$ n $\rightarrow$ I $\Rightarrow$ nt 'learn')						

Phonological rules 'Impugnment' Non-local and suprasegmental contexts

# Diagnosis

• 
$$\emptyset \to t / X \begin{bmatrix} -syl \\ 0-2son \\ 0-1aper \end{bmatrix} \_$$
 # covers two distinct sets of words  
=  $p,k,t,f,\theta,s,f,t' = b,d,g,v,\delta,z,g,d,m,n,\eta$ 

- Regular affixation after voiceless obstruents
- Irregular forms learnt, burnt, etc.
- Comparison of [-voi] and *n* should reveal that reliability is very different in these two contexts

• 
$$\emptyset \to t / \begin{bmatrix} -syl \\ 0-2son \\ 0-1aper \\ +voi \end{bmatrix} = p,k,t,f,\theta,s,f,f$$
  
•  $\emptyset \to t / \begin{bmatrix} -syl \\ 0-2son \\ 0-2son \\ 0-1aper \\ -voi \end{bmatrix} = b,d,g,v,\delta,z,g,dg,m,n,g$ 

### Impugnment

- When generalizing from rule R to superset R'
  - Compare confidence of subset context (R) and context outside subset (R'-R)
  - Be generous: lower confidence limit of subset, upper confidence limit of superset
- Impugnment: if superset confidence is lower, replace score of rule with (upper) confidence of superset region
- $\bullet~\mbox{Result:}~\ensuremath{\varnothing} \rightarrow \mbox{t}$  rule is adjusted from .528 to .090

Phonological rules 'Impugnment' Non-local and suprasegmental contexts

Image: A mathematical states and a mathem

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# Results of impugnment

Outputs for [spl1ŋ] with impugnment:

Output	Rule	Rel.	Conf.			
splīŋd		50/65	.700			
splæŋ	${\tt I} \rightarrow {\tt \varpi} \: / \: X \mathrel{{}^{\scriptscriptstyle }} \_\_ \{{\tt m,n,n}\}$	2/4	.310			
splɪŋt	$\varnothing \rightarrow t / X \begin{bmatrix} -syl \\ 0.2son \\ 0.1aper \end{bmatrix} $	79/142	.090			
$= p,k,t,f,\theta,s,f,tf$						
b,d,g,v,ð,z,ʒ,dʒ,m,n,ŋ						

Effect is stronger with larger training set

Phonological rules 'Impugnment' Non-local and suprasegmental contexts

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Image: A matrix and a matrix

### Some limitations of the model

- Input representations are 'flat'
  - No prosodic information (monosyllabic vs. polysyllabic)
  - No morphological information (inflection class, gender, etc.)
- Contexts are strictly local
  - No harmony, other non-local conditioning

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# Simulating tiers in a flat representation

Goal:

- Give the model access to prosodic size of the base
- E.g., monosyllabic vs. longer

Approach

- Create dummy segments, specified for just for 'length' feature
  - E.g., M (monosyllabic) vs. P (polysyllabic)
  - Feature: [±polysyllabic]

	Polysyllabic		
М	0		
Р	1		

• Could likewise specify gender/inflection class, nearest vowel, etc.

Phonological rules 'Impugnment' Non-local and suprasegmental contexts

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## Augmenting inputs with dummy segments

Simple case: all changes are suffixal (or prefixal)

• Use 'empty space' on other side of affix to specify suprasegmental context

lekleb <mark>P</mark>	leklebi <mark>P</mark>	fegrivP	fegrivi <mark>P</mark>
globlef <mark>P</mark>	globlef <mark>iP</mark>	pom <mark>M</mark>	pomi <mark>M</mark>
nuvM	nuvi <mark>M</mark>	sivodP	sivodi <mark>P</mark>
gor <mark>M</mark>	gori <mark>M</mark>	sogM	sogi <mark>M</mark>
fabeg <mark>P</mark>	fabegi <mark>P</mark>	zibotP	ziboti <mark>P</mark>

- Resulting rule format
  - ${\it \oslash} \rightarrow {\it i}$  / segmental context \_\_\_\_ prosodic context
- In practice, adding symbols to mark boundary can aid legibility
  - lekleb●P ~ leklebi○P, sog●M ~ sogi○M, ...
  - Change: ●→i○ (different input/output symbols ensure that they are not treated as part of the context)

# Discontinuous morphological contexts

- A particularly strong context for I→∧ (Bybee and Slobin 1982; Bybee and Moder 1983)
  - sC(C)\_\_\_\_ ŋ(k)
- Model presented here cannot discover this, due to its locality restriction
  - Alignment outward from change can't skip segments, or find features of multiple segments
  - Also cannot posit rules with literal segment outside featurally underspecified segments (sCC)
- Bybee and Moder (1983): speakers do generalize to novel sC(C)...items more than C(C)

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# A better alignment procedure

Albright and Hayes (2006): minimum string edit distance

- Aligns strings by finding optimal segmental correspondence between segments
- Align matching segments with one another
- Mismatches: find phonetically closest correspondent



# A challenge...

- Discontinuous alignments retain much more information than the 'myopic' alignments of the local MGL
- Explosion in number of rules
- Albright and Hayes (2006) devise a procedure for pruning the grammar to retain the most reliable rules, using the Gradual Learning Algorithm (Boersma 1997)
- Possible to employ more sophisticated selection/weighting techniques (e.g., maximum entropy models)
- Interesting theoretical issue: what kinds of non-local contexts do human learners actually notice?
  - English *v* initial verbs are all regular, but it appears that no effect of this is seen in wug verbs
- If you are interested in employing a non-local version of the model, let me know

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# Links

- Model download: http://www.mit.edu/ albright/mgl/
- These slides: http://www.mit.edu/ albright/mgl/MGL-Tutorial.pdf
- Additional papers, data files, etc.: http://www.linguistics.ucla.edu/people/hayes/learning/

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