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Your
Thinking*



2022 ASHA CONVENTION

Resilience REINVENTED

New Orleans • November 17-19
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Recognition of Aphasic Speech: ASR Development and Analysis

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National Institute on
Deafness and Other
Communication Disorders

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Presentation Overview

1. ASR for Clinical Assessment of Anomia
2. Post-Stroke Speech Transcription Challenge
3. ASR Analysis Tool: PhonoLogic Viewer
 - Download: <https://psst.study/phonologic/>
4. Q&A and Discussion

ASR for Clinical Assessment

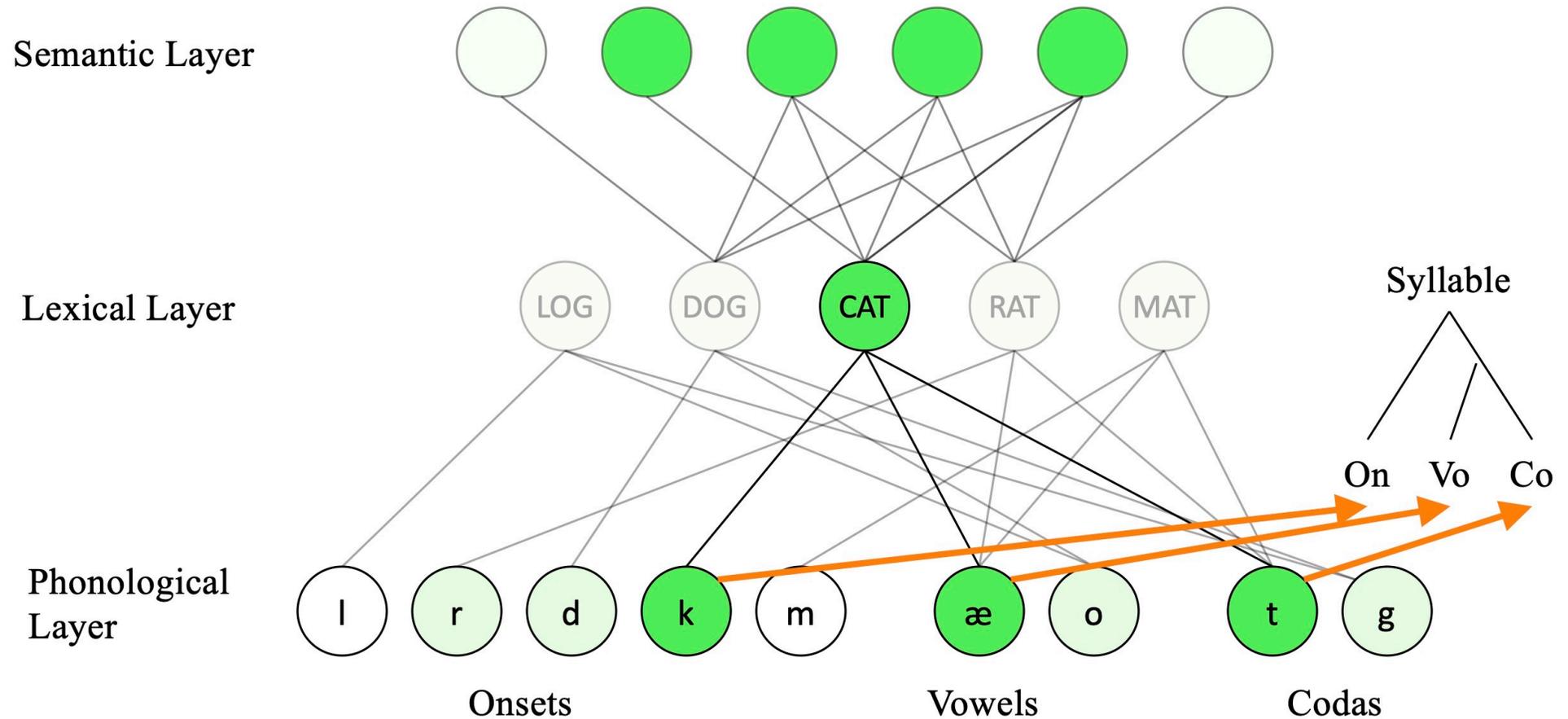
Who? people with aphasia

What? anomia

How? picture naming tests

Typical vs. Impaired Word Production

Dell's Model (Dell, 1986)



Anomia Assessment: Error Types and Analysis

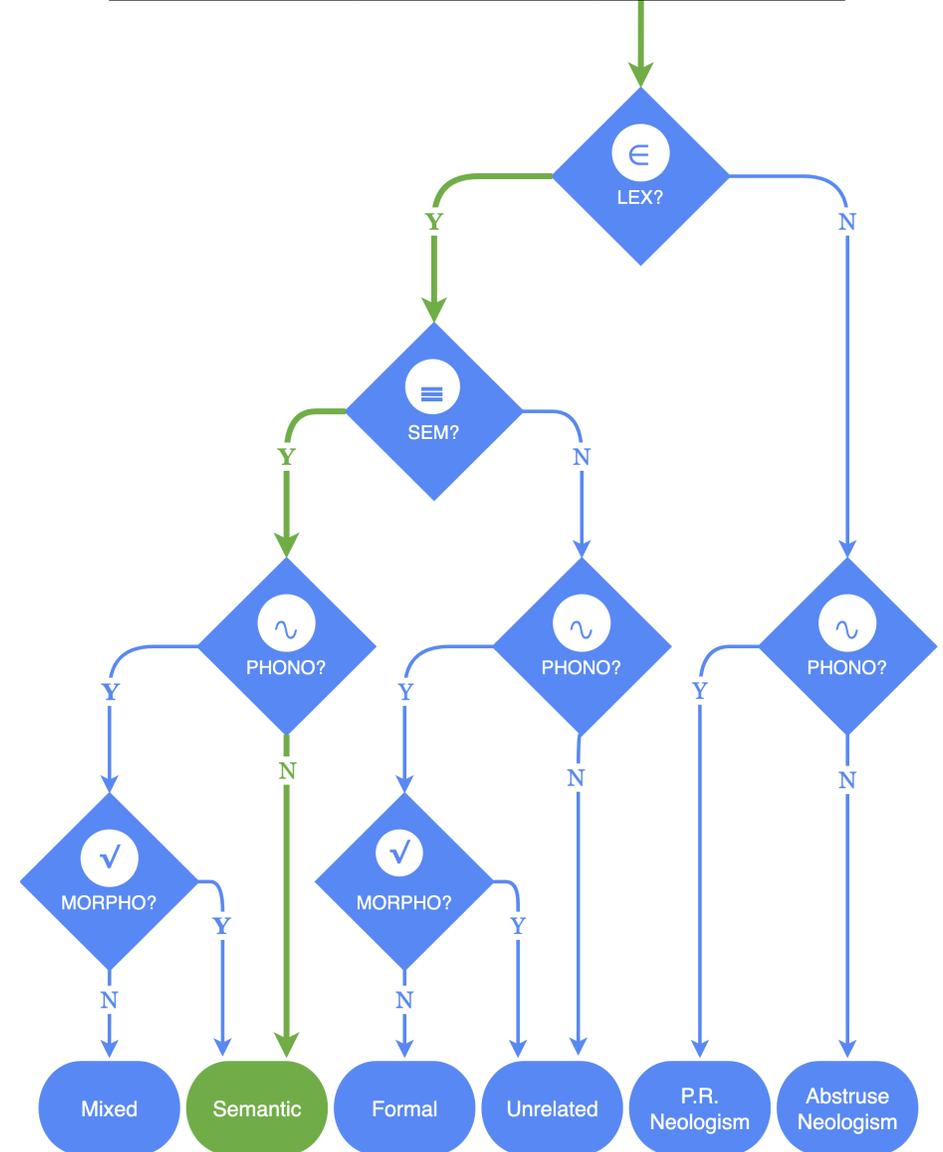
		Paraphasia Type	Features			Example
			Lexical	Semantic	Phonological	
Lexical		Semantic	+	+	-	"dog" for "cat"
		Formal	+	-	+	"cot" for "cat"
		Mixed	+	+	+	"rat" for "cat"
		Unrelated	+	-	-	"mug" for "cat"
Non-Lexical		Neologism	-	n/a	+	"tat" for "cat"
		Abstruse Neologism	-	n/a	-	"vop" for "cat"

Anomia Assessment: The Value of Automation

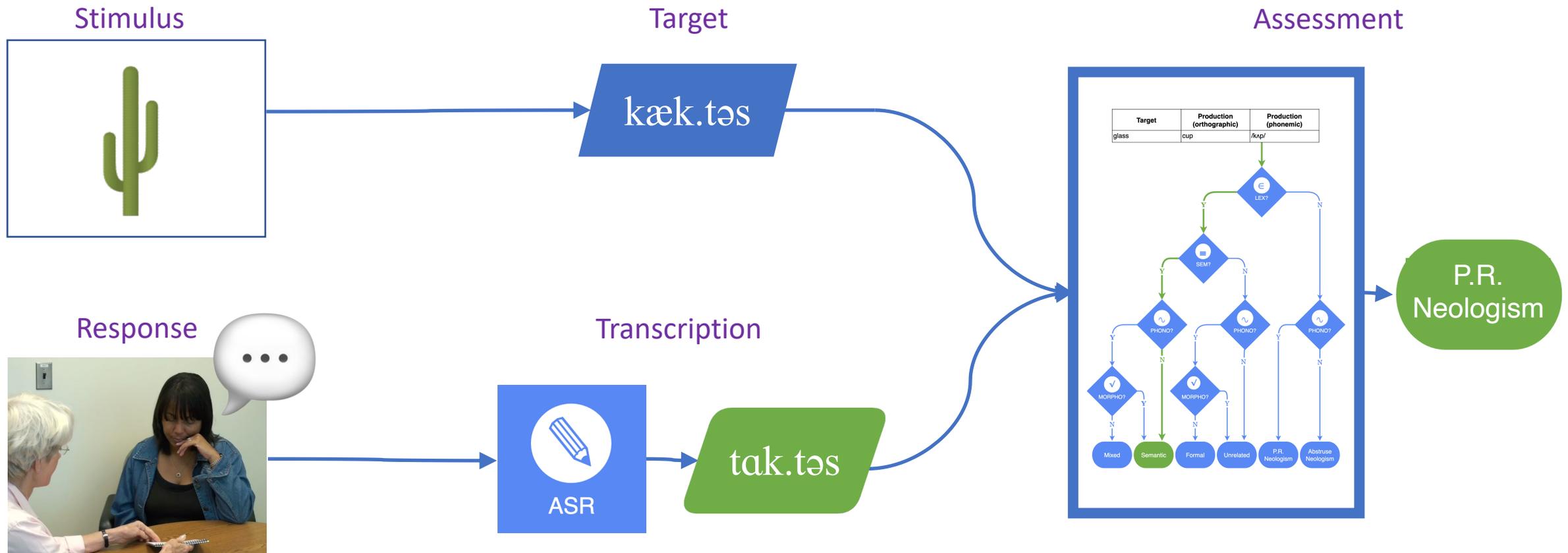


Algorithmic Classification of Paraphasias
aka “ParAlg” (Fergadiotis et al., 2016)

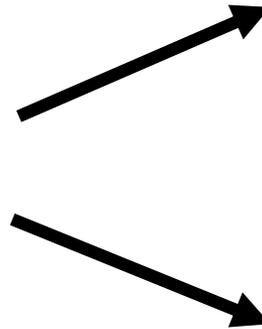
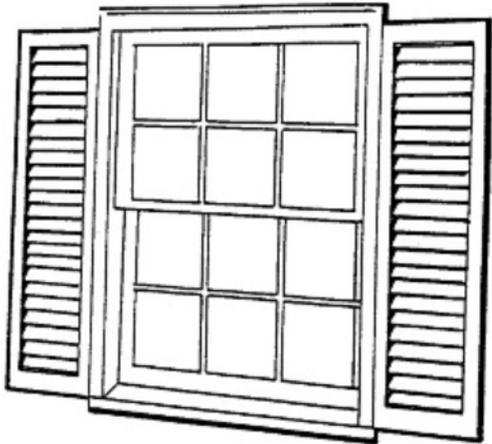
Target	Production (orthographic)	Production (phonemic)
glass	cup	/kʌp/



The Broader Vision: Fully Automated Anomia Assessment



ASR: Commercial vs. Clinical

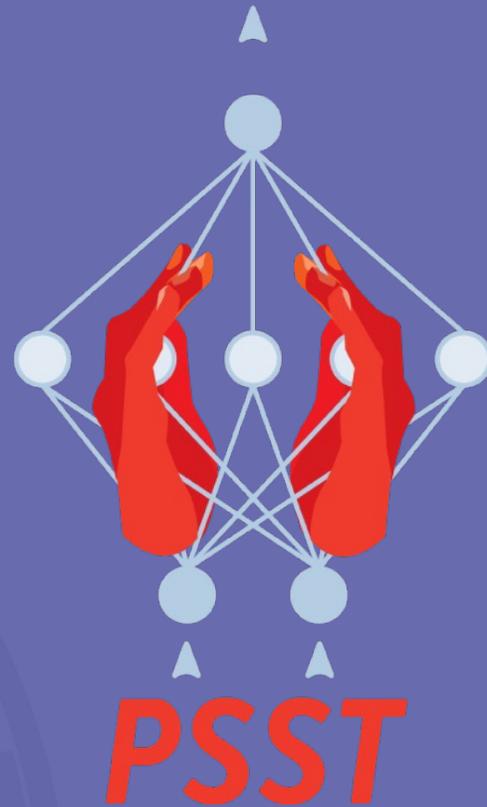


Commercial ASR:
/window/

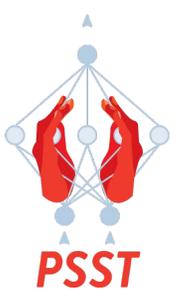
Clinical ASR:
[bindoʊ]

Post-Stroke Speech Transcription (PSST) Challenge

LREC 2022
Marseille



(Gale et al., 2022)



The PSST Challenge

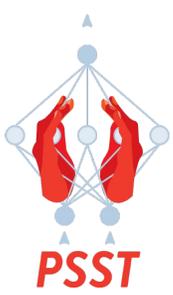
We provided

- A new dataset
 - Audio from English AphasiaBank (MacWhinney et al. 2011)
 - New phonemic transcripts
- A baseline phonemic ASR model
 - 26.4% phoneme error rate (PER)
 - 12.1% feature error rate (FER)

Challengers brought

- Clever new ideas
 - Several approaches to data augmentation
- An improvement on our baseline!
 - 20.0% phoneme error rate (PER)
 - 9.9% feature error rate (FER)

Gale et al. (2022) – <https://aclanthology.org/2022.rapid-1.6/>



PSST Speech Recognition Results

Yuan et al. (2022) →

Moëll/O'Regan.
et al. (2022) →

Our baseline →

Model	Arch	Data (hours of audio)					ASR	
		Pretrain	PSST	TIMIT	AphasiaBank	Other	FER	PER
Y1	LARGE	60,000	2.8		33.3 ^U		9.9%	20.0%
Y2	LARGE	60,000	2.8	3.9			10.3%	21.1%
Y3	LARGE	60,000	2.8		44.0 ^W		10.4%	21.5%
Y4	LARGE	60,000	2.8			3.9 ^L	10.6%	22.2%
Y5	LARGE	60,000	2.8				10.9%	22.3%
MO1	LARGE	960	2.8	1.1 ^r			11.3%	25.5%
MO2	LARGE	960	5.6 ^p				11.4%	25.1%
MO3	BASE	960	2.8	1.1 ^r			11.7%	26.3%
MO4	LARGE	960	5.6 ^t				11.7%	25.4%
MO5	LARGE	960	5.6 ^p	1.1 ^r			11.9%	26.0%
MO6	LARGE	960	2.8				12.0%	25.9%
MO7	BASE	960	5.6 ⁿ				12.0%	26.1%
<i>PSST-A</i>	BASE	960	2.8				12.1%	26.4%
Y6	LARGE	60,000	2.8			100 ^L	12.5%	26.0%
Y7	LARGE	60,000	2.8			960 ^L	16.7%	38.0%

^L Librispeech, pseudo-labeled with G2P
^U iteratively pseudo-labeled (unweighted)
^W iteratively pseudo-labeled (weighted)

^p with pitch-shifted variants
^t with time-shifted variants
ⁿ with Gaussian noise augmentation

^r RIR reverb applied

Evaluating an ASR

Word error rate (WER)

Orthographic ASR: $\frac{\# \text{ WORD ERRORS }}{\# \text{ TARGET WORDS }}$

Human: a house



ASR: a horse



$$\frac{1}{2} = 50\% \text{ WER}$$

Phoneme Error Rate (PER)

Phonemic ASR: $\frac{\# \text{ PHONEME ERRORS }}{\# \text{ TARGET PHONEMES }}$

Human: t a k t ə s



ASR: d a k t ə s



$$\frac{1}{6} = 17\% \text{ PER}$$

Further intuition: /taktəs/ → /daktəs / should score better than /taktəs/ → /oaktəs/

Phonological Features

p = <voiceless>	<bilabial>	<stop>
b = <voiced>	<bilabial>	<stop>
t = <voiceless>	<alveolar>	<stop>
d = <voiced>	<alveolar>	<stop>
k = <voiceless>	<velar>	<stop>
g = <voiced>	<velar>	<stop>

ARPAbet	IPA	consonantal	delayedrelease	continuant	sonorant	approximant	syllabic	tap	nasal	voice	spreadglottis	labial	round	labiodental	coronal	anterior	distributed	strident	lateral	dorsal
P	p	+	-	-	-	-	-	-	-	-	-	+	-	-	-	0	0	0	-	-
B	b	+	-	-	-	-	-	-	-	+	-	+	-	-	-	0	0	0	-	-
T	t	+	-	-	-	-	-	-	-	-	-	-	-	-	+	+	-	-	-	-
D	d	+	-	-	-	-	-	-	-	+	-	-	-	-	+	+	-	-	-	-
K	k	+	-	-	-	-	-	-	-	-	-	-	-	-	-	0	0	0	-	+
G	g	+	-	-	-	-	-	-	-	+	-	-	-	-	-	0	0	0	-	+

Distance between two phonemes

- Feature system: a table of distinctive features
 - Modified version of Hayes (2009)
 - 24 features x 40 phonemes
- Consider each phoneme as a set of features
- Error cost as a vector distance:

$$\text{Cost}(s, f) = \|\vec{s} - \vec{f}\| = \left\| \begin{bmatrix} +\text{cons} \\ +\text{delrel} \\ +\text{cont} \\ +\text{ant} \\ -\text{dist} \\ \dots \\ -\text{voi} \end{bmatrix} - \begin{bmatrix} +\text{cons} \\ +\text{delrel} \\ +\text{cont} \\ -\text{ant} \\ +\text{dist} \\ \dots \\ -\text{voi} \end{bmatrix} \right\| = \left\| \begin{bmatrix} 0 \\ 0 \\ 0 \\ -1 \\ 1 \\ \dots \\ 0 \end{bmatrix} \right\| = 2 \text{ features apart}$$

Special considerations (don't worry too much about these)

- Values can be:
 - Present [+]
 - Absent [−] or
 - Not relevant [0]
- Diphthongs
 - Calculate as one phoneme or two?
 - Workaround, new values:
 - Absent-to-present [−+]
 - Present-to-absent [+−]

Cost	Feature Changes		
1	[−feature]	↔	[+feature]
0.75	[−feature]	↔	[+−feature]
	[−+feature]	↔	[+feature]
0.5	[−feature]	↔	[0feature]
	[−+feature]	↔	[+−feature]
	[0feature]	↔	[+feature]
0.25	[−feature]	↔	[−+feature]
	[−+feature]	↔	[0feature]
	[0feature]	↔	[+−feature]
	[+−feature]	↔	[+feature]
0	[−feature]	↔	[−feature]
	[−+feature]	↔	[−+feature]
	[0feature]	↔	[0feature]
	[+−feature]	↔	[+−feature]
	[+feature]	↔	[+feature]

Distance between two *transcripts*

- Similar to PanPhon (Mortensen, 2016)
- Find alignment with least error (Levenshtein, 1966)
- Insertions & deletions: ignore undefined features

Phoneme Error Rate (PER)		vs.	Feature Error Rate (FER)	
<i>Human:</i>	l æ f ɪ n		<i>Human:</i>	l æ f ɪ n
<i>ASR:</i>	b ɹ a p ɹ ɪ ŋ		<i>ASR:</i>	b ɹ a p ɹ ɪ ŋ
	✗ ✗ ✗ ✗ ✗ ✓ ✗			
	$= \frac{6}{5} = 120\%$		$\frac{22}{24} \frac{4}{24} \frac{2}{24} \frac{3}{24} \frac{23}{24} \checkmark \frac{23}{24} = \frac{58.5}{130} = 45\%$	

Feature distance sounds very promising, but...

- Even when you understand the principles...
 - Unreasonable to estimate in your head
- Even when you're looking at the answer...
 - Difficult to explain why
- Cross-disciplinary: linguistics, computer science
- Cumbersome: dozens of features per phoneme, alignment

Don't fret, though...

PhonoLogic Viewer

an ASR analysis tool

Download:
<https://psst.study/phonologic/>

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Questions?

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