Judgements Gradient e Word-Edge Syllable

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Motivation

- How are word-medial syllable boundaries learned?
- Simplifying work: boundaries from by Maximum Onset Principle. assumption ₽. previous

additional influences. Empirical data: variation and

> segmentation: to statistics sensitivity Word

syllabification sensitive lexical statistics from word edges? to gradient, joint

S

Maximum Onset Principle(MOP)

- ment Word-medial syllable boundary place onset (Kahn 1976) to yield longest possible legal
- kaep.stan, not kaeps.tan

Where are syllable boundaries?

- Variable and non-MOP medial syllable boundary judgements:
- (Content 2012), Polish (Rubach & Booij 1990), dington et al. 2013) Czech(Šturm 2018), English (Edet al. 2001), lrish (Chiosain et French
- Eddington et al. (2013):
- ~5,000 English lexical items
- Each presented to ~25 participants
- ment Forced choice syllabification judge

Joint Word Edge Score

- Frequency nunciation Dictionary estimates from CMU Pro-
- P(Onset): estimated from word-initial
- P(Coda): estimated from word-final

JointWordEdgeScore of $P(Onset) \times P(Coda)$ a syllabification:

JointWordEdgeScore of kaep.stan P(Onset st) \times P(Coda

- Normalized for each lexical item
- Accounts for coda

Hypothesis

- Assuming analyzing Eddington tal data. et <u>a</u> 2013 model experimen
- Including k p edictors of syllabificatio
- then: If English s tations are sensitive speakers' to syllable joint wordboundary edge statistics repres
- dictor w Includin $\stackrel{\sim}{=}$ the significantly joint Wor impro edge mo atis del

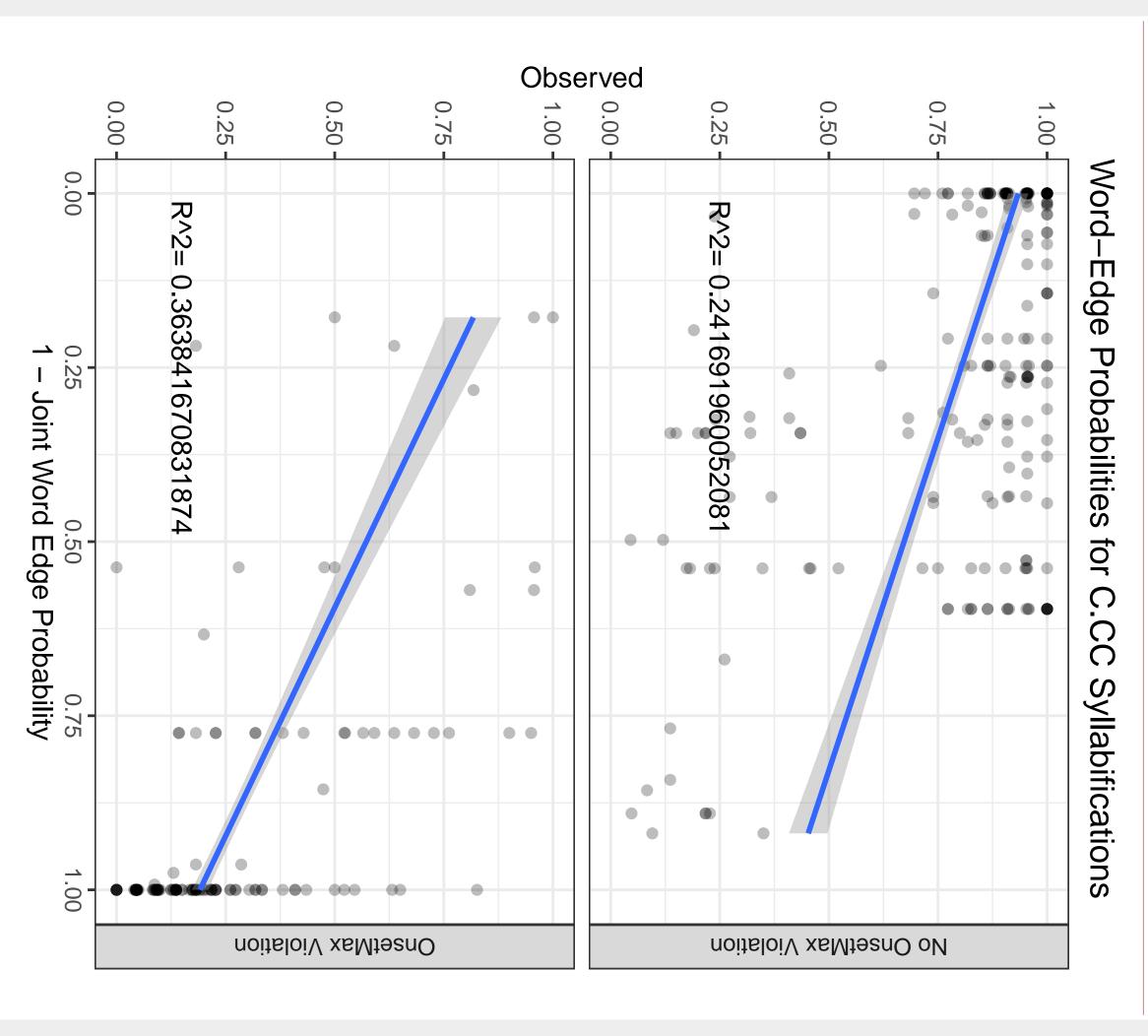
Results: Z del Comparison

Model Predictors	Log Likelihood	BIC	∆BIC
Morph, Stress,	E1122 2	100200 0	1 5 5 7 1
OnsetMax			T323.1
Morph, Stress,			
OnsetMax,	-50365.01	100775.8 0	0
JointWordEdgeScore			

Table: Model compariso inclusion of word-edge s statistics (the of multinomial regress JointWordEdgeSc ion with and without edictor).

Bayes Factor of Joint Word Edge Score100: model decisive evidence for

Results: Joint Word Edge Scor D Distribution



- one lexical Each point the (with syllabifica medial ation option
- Vertical: how labification Joint Word Edge sponse. often pa \bigcirc ticipants the <u>..</u> andidate chose
- Top: candi dates obeying MOP

scussion

- influence wordfication initial English and word-final statistics pothesis word-medial syllabithat joint
- statistics imilar zech und for Šturm (2018) word-initial
- Sa entations sible and learning: ons for syllable rep-
- bootstr de. boundaries? learr ning of word-medial word boundarie
- boundaries e18 word-medial dient? syllable
- ф. implicit syllable asures boundaries? of speakers'
- Correlation between MOP and lexical
- models Incorporating joint <u>of</u> syllable segmentation word edge scores in learn-

onclusion

- final Simple lexical measur statistics vord-initial scoring syllabifica and word-
- speaker statistics, del beyond syllabif mparison MOP onset, ications shows coda partly English lexical e X

References

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