

Interaction of high vowel devoicing and syllabification

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Introduction

What we would like to do

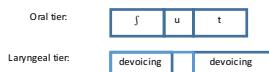
- Japanese high vowels are “devoiced” in some environments.
- Q1: Are these vowels deleted? Or are they devoiced?
- Q2: If they are deleted, then what’s the consequence for morafication/syllabification?
- (Q3: if Japanese speakers have CVC syllables due to high vowel deletion, are they sensitive to syllable-related phonotactic restrictions, possibly provided by UG: cf. Berent’s recent work?)
- We offer new articulatory data on the table to bear on these questions.

Japanese high vowel devoicing

- The standard description: High vowels are devoiced between two voiceless consonants.
- An example, [ʃutaisei] “willingness”.
- “Devoicing” is the traditional term used. However, whether these vowels are simply devoiced or deleted still remain debated.
 - Devoiced=oral gestures remain intact or reduced.
 - Deleted=oral gestures deleted.

Devoicing vs. Deletion?

- A phonetic devoicing view:
 - Laryngeal gestures overlap to passively devoice vowels (e.g., Jun & Beckman 1993).



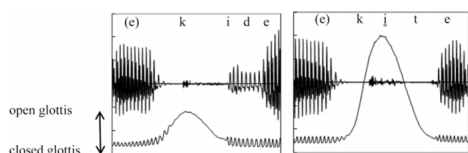
- A phonological deletion view:
 - $V \rightarrow \emptyset / [-\text{voice}] _____ [-\text{voice}]$ (Kondo 2000)
 - Or a phonological process induced by a set of markedness constraints (Tsuchida 1997).

Or could it be both?

- Kawakami (1971) lists environments where the target vowels are deleted and where they are devoiced.
- But Kawakami gives no instrumental evidence.
- Whang (2014, 2016) argues that vowels are deleted when their identity can be recovered from the preceding consonants.
- For example, the devoiced vowel is predictably [u] after [ɸ], but it is not after [k]. Only in the former environment can the vowel be deleted.

Japanese laryngeal gestures

- In devoiced vowel contexts, e.g., /**kite**/, there is a single laryngeal gesture of **greater magnitude** than a single consonant gesture, c.f. /**kide**/ (Fujimoto et al. 2002)



How about lingual gestures?

- However, the lingual gesture of devoiced vowels is understudied.
- Funatsu & Fujimoto (2011), as far as we know, is the only exception. They used EMMA, EGG, and PGG.
- Their EMMA results show only small differences between voiced [i] and devoiced [i] (implying “devoicing”, not “deletion”).
- But they studied only one speaker. And only one pair of stimuli ([kite] vs. [kide]), one of whose member is a real word. 4 repetitions and no quantitative analyses.

Summary of past work

- The laryngeal data indicate that devoicing in Japanese is actively controlled.
- Acoustic evidence for presence/absence of a lingual vowel articulation has been largely equivocal, with some studies claiming the vowel has been deleted, while others claiming that it is present.
- Deletion: Beckman 1982; Beckman & Shoji 1984; Kondo 1997, 2000.
- Devoicing: Jun & Beckman 1993; Faber & Vance 2001; McCawley 1968.
- Both possible: Kawakami 1971; Tsuchida 1997; Whang 2014.

Our contribution to the debate

- Electromagnetic Articulography (EMA) data on the **lingual articulation** of vowels in voiced and devoiced contexts.
- A **computational method** for evaluating vowel deletion on the basis of movement kinematics (Shaw & Kawahara, submitted).
- Showing that some cases do involve cases of actual **deletion**.
- Some implications for prosodic organization of Japanese (still in progress).

EMA Experiment

EMA experiment

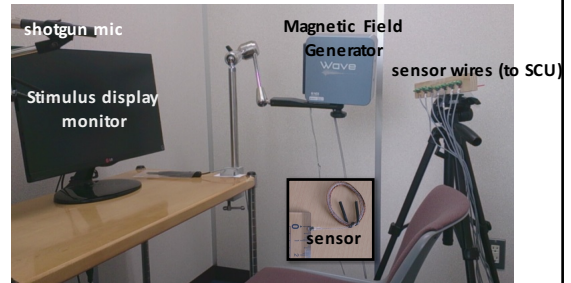
- Record the articulatory dynamics of voiced vowels and devoiced vowel counterparts.
- Apply Discrete Cosine Transform to fit the observed data.
- Use a Bayesian classifier to evaluate the likelihood that devoiced tokens have a lingual articulation.

Materials (target words)

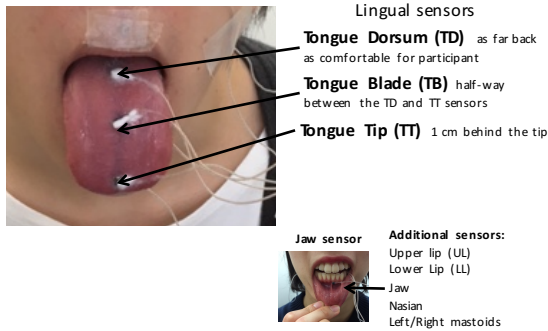
	Devoicing/deletion	Voiced vowel
Entropy = 1.99 (w, ㄹ)	mas <u>u</u> taa マスター	mas <u>u</u> da 益田
Entropy = 1.89 (w)	hak <u>u</u> sai 白菜	yak <u>u</u> zai 薬剤
Entropy = 1.46	j <u>u</u> tasei 主体性	j <u>u</u> daika 主題歌
Entropy = 1.08 (w, ㄹ)	φ <u>u</u> soku 不足	φ <u>u</u> zoku 付属
Entropy = 0.09 (w, ㄹ)	kats <u>u</u> toki 勝つ時	kats <u>u</u> dou 活動

10-15 repetitions of the target words in the carrier phrase: oke _____ to itte 'Ok, say _____'. Participants were instructed to speak as if they were making a request of a friend. Interspersed with 10 words lacking /u/

Apparatus



Sensor placement



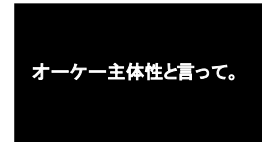
Procedure

- Stimuli were presented on a monitor in random order.
- To promote fluent reading, target words were previewed before displayed in the carrier sentence.
- A native speaker of Japanese monitored pronunciation manually advancing trials after accepting or rejecting each token.

Target preview (500ms)

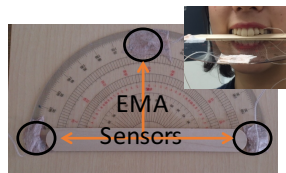
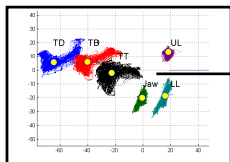


Target sentence



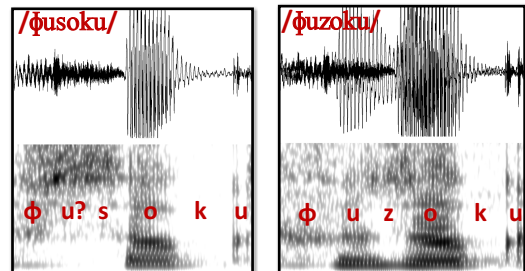
Post-processing

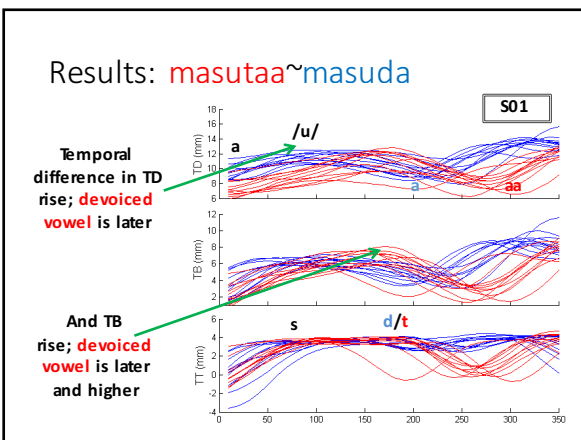
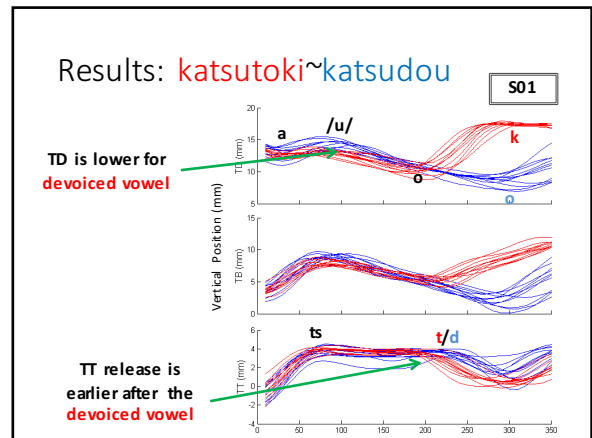
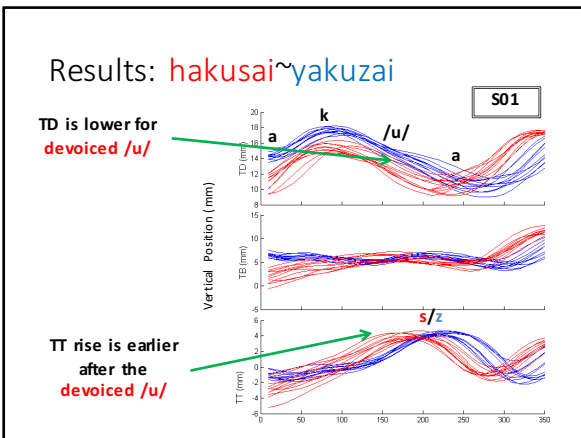
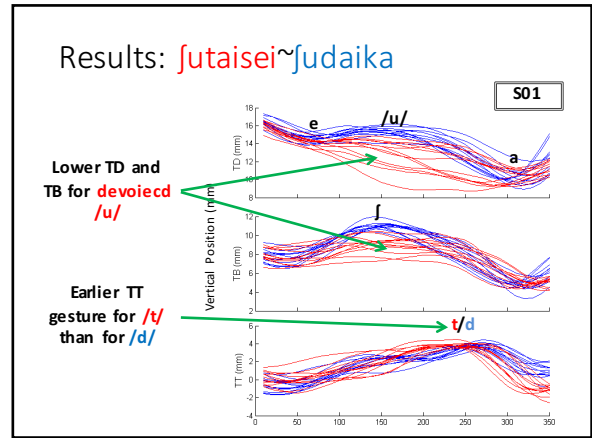
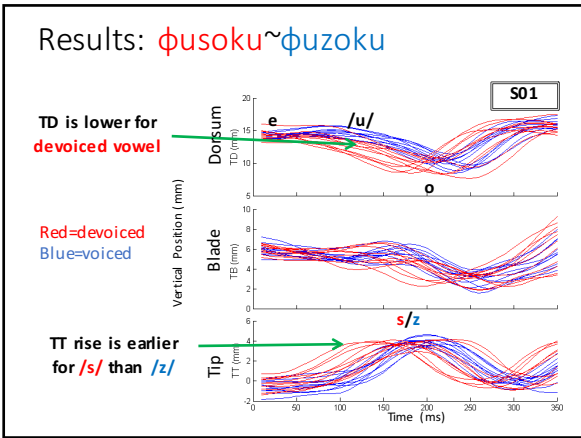
- Head movements corrected computationally
- Data rotated to the occlusal (bite) plane
- Robust smoothing (Garcia 2010)



Acoustic results: vowel devoicing

In line with current descriptions of Tokyo Japanese, /u/ was **devoiced** between voiceless consonants and voiced otherwise.





Computational analysis

Analysis

- **Voicing effect:** do voiced and devoiced vowel trajectories differ?
- **Phonological deletion:** is the vowel /u/ absent in any of these words?
 - Discrete Cosine Transform (DCT) on TD trajectories
 - Simulate a "targetless" trajectory
 - Compare DCT coefficients of "targetless" trajectory to data, c.f., t-test against zero.

Discrete Cosine Transform (DCT)

Complex curve represented as the sum of Cosines:

$$C(m) \cos(\theta)$$

where m is the n th cosine and θ is a function of length

$$C(m) = \frac{2}{N} \sum_{n=0}^{N-1} x(n) \cos\left(\frac{(2n+1)(m-1)\pi}{2N}\right)$$

Where N is the number of data samples;
 $M = 1, \dots, N$; $k_m = \frac{1}{\sqrt{2}}$ when $m = 1$, else $k_m = 1$;
 $x(n)$ is the intercept;

How many DCT coefficients are needed?

- Nearly lossless compression (99.6%) with 6 coefficients.
- We used 4 DCT coefficients (99.0%)

Each cosine components

- Raw data (green)
- Mean DCT (black)
- [e]-to-[a] line (red)

- 1st DCT Coefficient → Average TD height
- 2nd DCT Coefficient → V-to-V trajectory
- 3rd DCT Coefficient → Intervening vowel
- 4th DCT Coefficient → Consonantal effects

Vowel deletion ("noisy null") trajectories

raw data (green lines)
 direct e-to-a trajectory (red line)

Noisy null trajectories (black lines) generated from stochastic sampling of Gaussian distributions defined by mean DCT coefficients fit to the direct e-to-a trajectory (red line) and the standard deviation of DCT coefficients fit to the raw data (green lines).

Within-speaker, within-word variation

Phonetic reduction
 Or variable deletion?

Almost all tokens look pretty different from noisy null

Noisy null

Some tokens look more like noisy null, whereas others look like they have a clear target.

Token-by-token evaluation

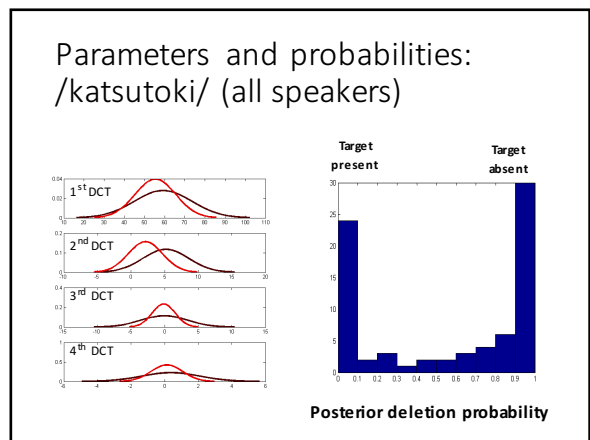
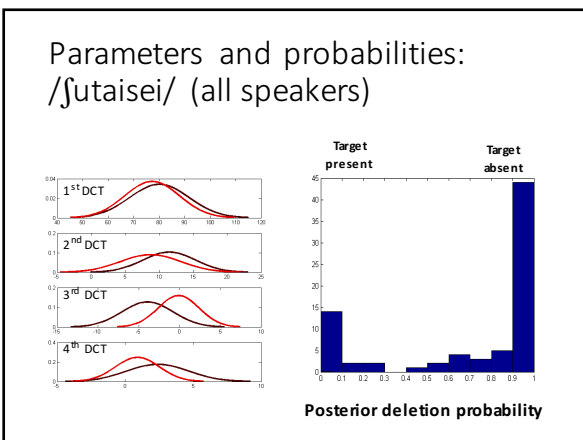
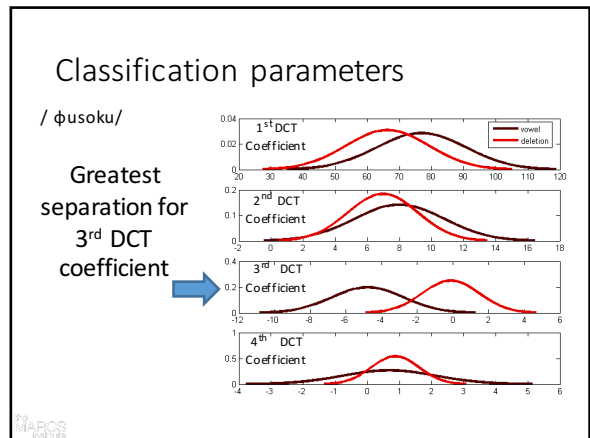
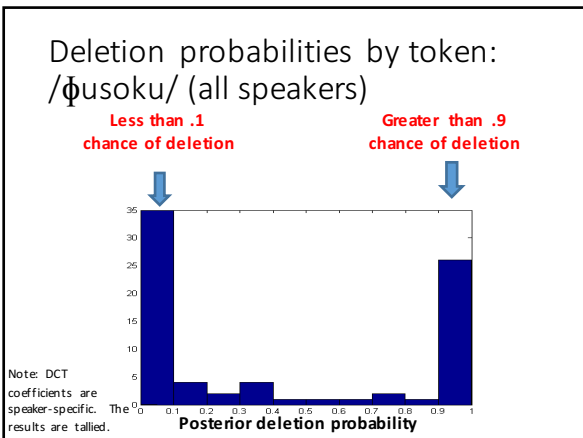
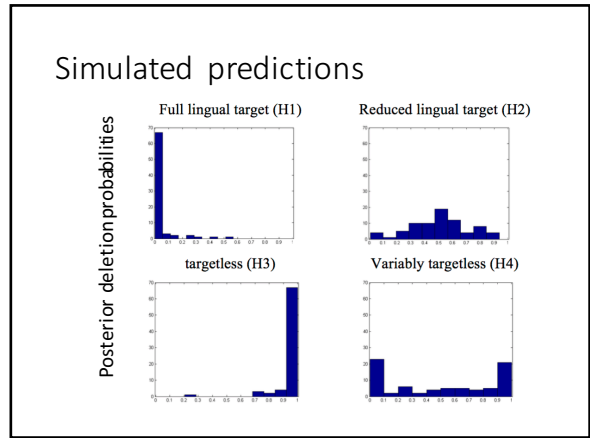
Fit a naïve Bayes classifier to the data and used it to generate (posterior) deletion probabilities

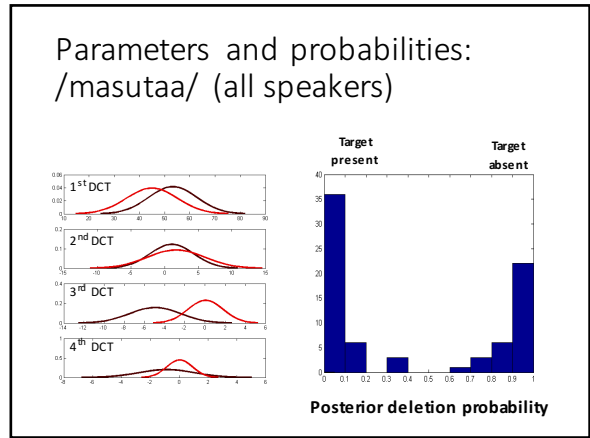
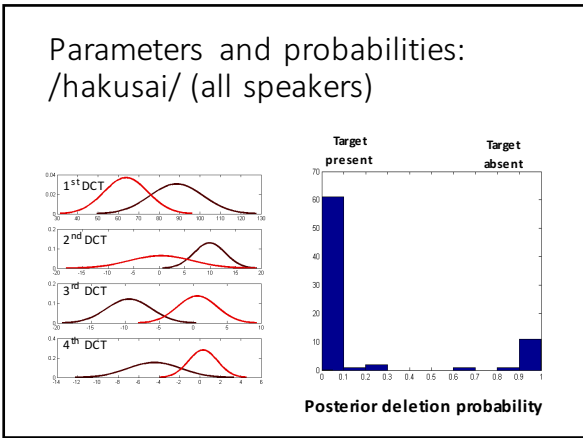
Training data = **voiced tokens** (full target) & **noisy null** (no target)
 Test data = **voiceless tokens**

$$p(D|c_1, \dots, c_d) = \frac{p(D) p(c_1, \dots, c_d|D)}{p(c_1, \dots, c_d)}$$

where...

- $D = \{deletion, full\ vowel\}$
- $c_1 = 1^{st}$ DCT Coefficient
- $c_2 = 2^{nd}$ DCT Coefficient
- $c_3 = 3^{rd}$ DCT Coefficient
- $c_4 = 4^{th}$ DCT Coefficient





All the results

	S01	S02	S03	S04	S05	S06	average
φ _{usoku}	0.47	0.39	0.75	0.84	0.01	0.19	0.44
ʃ _{tasei}	0.92	0.68	0.84	0.99	0.02	0.89	0.72
kats _{utoki}	0.81	0.19	0.69	0.93	0.06	0.79	0.58
hak _{usai}	0.00	0.00	0.51	0.50	0.00	0.07	0.18
mas _{utaa}	0.64	0.09	0.01	0.02	0.74	0.41	0.32
average	0.57	0.27	0.56	0.66	0.17	0.47	0.45

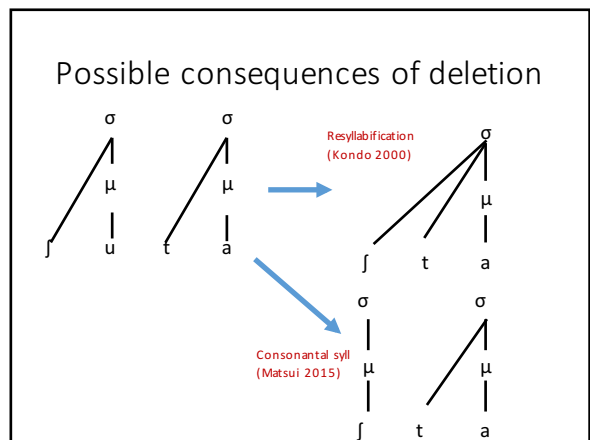
- /ʃ_{tasei}/ shows the highest probability of deletion; /hak_{usai}/ the lowest.
- S03 & S04 show high deletion probabilities; S05 barely showed deletion.

Effect of surface consonant cluster type?

- All the histograms are bi-modal, supporting the optional deletion hypothesis.
- For all speakers we find:

ʃ_t >> φ_s >> k_s
 ↙ ↘
 ts_t

Implications for phonological organization



Mora remains

- A bimoraic truncation pattern (Poser 1980 *et seq.*) counts moras of a “devoiced” vowel as moraic (Kawahara 2015; Tsuchida 1997).
- E.g. [suto] < [sutoraiki] (loanword truncation)
- E.g. [tʃjka(-tʃan)] < [tʃikako] (hypocoristic)
- E.g. [ɸuka-ɸuka] (mimetics)
- Hirayama (2009) shows that devoiced vowels count as much as voiced vowels for *haiku*.

Syllable remains too (?) Ito (1990), Kawahara (2016)

Bimoraic truncation patterns

- [de.mon.su.to.ree.ɸon] → [de.mo] ‘demonstration’
- [ri.haa.sa.ru] → [ri.ha] ‘rehearsal’
- [ro.kee.ɸon] → [ro.ke] ‘location’

Monosyllabic outputs are not allowed: a light syllable is appended

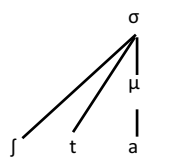
- [dai.ja.mon.do] → [dai.ja], *[dai] ‘diamond’ = a word must be bisyllabic; Wd much branch (I&M 1992)
- [paa.ma.nen.to] → [paa.ma], *[paa] ‘permanent (hair style)’
- [kom.bi.nee.ɸon] → [kom.bi], *[kon] ‘combination’
- [ɸim.po.ɸi.u.mu] → [ɸim.po], *[ɸin] ‘Symposium’

Devoiced vowels count = a devoiced vowel projects its syllable

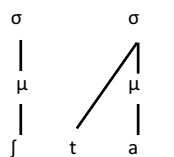
- [mai.ku.ro.ɸo.on.] → [mai.ku], *[mai] ‘microphone’
- [am.pu.ri.ɸai.aa] → [am.pu], *[an] ‘amplifier’

Syllable remains too (?)

- Predictions from the cross-linguistic perspectives:



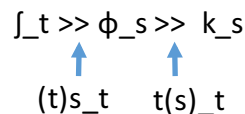
Rising sonority is better; a.k.a. Sonority Sequencing Principle (SSR)



Falling sonority is better; a.k.a. Syllable Contact Law (SCL)

Effect of surface consonant cluster type?

- The hierarchy we found supports the second view; i.e. that two consonants are separated by a syllable boundary.



- Affricates can variably be treated as an ordered fricative-stop segmental complex (Sagey 1986) or strident stop (Clements 1999) (cf Lombardi 1990).

A caveat

- Each phonological environment was tested with one item only.
- A follow-up experiment has been run with an additional 6 speakers producing each of the below dyads 10-15 times (but the data is yet to be analyzed).

FS	FF	SS	SF
ɸuton~ɸudou (布団—不働)	ɸusoku~ɸuzoku (不足—付属)	kutakuta~kudaranu <た<た-<たらぬ	kusami~kuzai 臭み—匂材
ɸutan~ɸudan (負担—不断)	ɸusa~ɸuzai (負債—不在)	kutabaruu~kudasaru くたばる くださる	kusari~kuzawa 鎖 久沢
ɸuta~ɸuda (ふた—礼)	ɸusagaru~ɸuzakeru ふさがる ぶざける	kutaniyaki~kudanshita 九谷焼 九段下	kusakar~kuzakicho 草刈り 久崎町
jutaisei~judaika (主体性—主題歌)	jusai~juzai (主催—取材)	Entropy also varied across items	
jubou~judou (酒造—手動)	jusa~juzan (主査—珠算)		
jutokou~judouken 首都高 主導権	juso~juzou 主訴 酒造		

Temporal stability analysis

- Cross-linguistic work on the articulatory timing of consonant clusters has identified timing differences correlated with syllable structure.

- These differences are reflected in patterns of temporal stability across CVX and CCVX sequences (Browman and Goldstein, 2007; Shaw et al., 2009; Marin, 2012; Hermes et al., 2013; Shaw and Gafos, 2012).

Patterns of temporal alignment

Heterosyllabic parse (simplex syllable onsets) Tautosyllabic parse (complex syllable onsets)

Syllable Parse [C.CV] [CCV]

On the hypothesis that the syllable nucleus is coordinated with the syllable onset... (Browman and Goldstein, 2000)

Coordination Topology

Surface Pattern

Temporal stability metrics

Following: Browman, C.P. Goldstein, L. (1988) Some Notes on Syllable Structure in Articulatory Phonology. *Phonetica* 45: 140-155. PMID: 3255974

Simplex Onset Alignment (Arabic) Complex Onset Alignment (English)

Relative Standard Deviation (RSD)		
left	center	right
.12	.07	.04

Relative Standard Deviation (RSD)		
left	center	right
.05	.02	.07

Singleton control words for stability analysis (already recorded)

	Consonant cluster	Singleton control
Entropy = 1.99 (w, ㄨ)	[mas ^h taa] マスター	ba ^h taa パター
Entropy = 1.89 (w)	[haksai] 白菜	da ^h sai ダサイ
Entropy = 1.46	[[t ^h aisei] 主体性	t ^h aisei 体制
Entropy = 1.08 (w, ㄨ)	[f ^h soku] 不足	ka ^h soku 加速
Entropy = 0.09 (w, ㄨ)	[kats ^h toki] 勝つ時	mi ^h ru ^h toki 見る時

Methods

Bayesian decision rule applied to posterior probabilities

Target absent tokens (n = 138) were compared to singleton controls

Posterior deletion probability

Preliminary results: futaisei

shutaisei: syllable-referential intervals

Relative Standard Deviation (RSD)		
LE_A	CC_A	RE_A
.23	.18	.11

The right-edge to anchor (RE_A) Interval is the most stable, an indication of simplex onsets

Data from a total of 276 tokens, 138 taisei and 138 shutaisei with <5 probability of vowel target; NB: data are from 5 speakers, as one speaker (S05) had no targetless /u/ tokens.

Conclusion

- The articulatory nature of devoiced vowels was barely known (*modulo* Funatsu & Fujimoto 2011).
- We provided first systematic analyses of the relevant articulatory data.
- Vowel is either **deleted** or **retained**, but **never (or only very rarely) reduced**.
- The likelihood of deletion varies across different consonantal environments, but systematically so.
- There is inter-speaker variation as well.
- Not entirely consistent with Kawakami's (1971) or Whang's (2014, 2016) predictions.
- Not clear effects of consonant-conditioned entropy either.

- We conjectured that our results support the heterosyllabic analysis of the "resulting clusters" (Matsui 2005).
 - Some phonological evidence
 - Syllable contact effect
 - Temporal stability patterns
- But more needs to be done, especially to confirm that Japanese follows Syllable Contact Law (the emergence of the unmarked?).

Acknowledgements

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- Thanks to Jeff Moore and Chika Takahashi for their help with the experiment and EMA analysis.