

Anterior-posterior distension of maximal upper esophageal sphincter opening is correlated with high-resolution cervical auscultation signal features

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Abstract.

Objective: Adequate upper esophageal sphincter (UES) opening is essential during swallowing to enable clearance of material into the digestive system, and videofluoroscopy(VF) is the most commonly deployed instrumental examination for assessment of UES opening. High-resolution cervical auscultation (HRCA) has been shown to be an effective, portable and cost-efficient screening tool for dysphagia with strong capabilities in non-invasively and accurately approximating manual measurements of VF images. In this study, we aimed to examine whether the HRCA signals are correlated to the manually measured **AP anterior-posterior (AP)** distension of maximal UES opening from VF recordings, under the hypothesis that they would be strongly associated. *Approach:* We developed a standardized method to spatially measure the AP distension of maximal UES opening in 203 swallows VF recording from 27 patients referred for VF due to suspected dysphagia. Statistical analysis was conducted to compare the manually measured AP distension of maximal UES opening from lateral plane VF images and features extracted from two sets of HRCA signal segments: whole swallow segments and segments excluding all events other than the duration of UES is opening. *Main results:* HRCA signal features were ~~statistically significantly associated to~~ significantly associated with the normalized AP distension of the maximal UES opening in the longer whole-swallowing segments and the association became much stronger when analysis was performed solely during the duration of UES opening. *Significance:* This preliminary feasibility study demonstrated the potential value of HRCA signals features in approximating the objective measurements of maximal UES AP distension and paves the way of developing HRCA to non-invasively and accurately predict human spatial measurement of VF kinematic events.

Keywords: high-resolution cervical auscultation, upper esophageal sphincter, swallowing accelerometry, swallowing sounds, signal processing, deglutition

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1. Introduction

Upper esophageal sphincter (UES) opening during swallowing is an important event facilitating the passage of ingested materials into the esophagus, and its dysfunction can lead to inefficient clearance. This in turn, contributes to aspiration of hypopharyngeal residue and associated adverse pulmonary complications (Jacob, Kahrilas, Logemann, Shah & Ha 1989, Kim, Park, Oommen & McCullough 2015, Singh & Hamdy 2005, Cook, Dodds, Dantas, Massey, Kern, Lang, Brasseur & Hogan 1989). Several physiologic events contribute to opening of the UES during swallowing, and the accurate measurement of the duration and anterior-posterior (AP) distension of UES opening is essential in estimating the contribution of impaired opening to these risks (Jacob et al. 1989, Kim et al. 2015).

The videofluoroscopic (VF) swallowing study is the only imaging-based swallowing assessment technique capable of enabling visualization of UES opening in order to assess its degree of AP distension during swallowing (Sejdić, Malandraki & Coyle 2019, Ahuja & Chan 2016, Martin-Harris & Jones 2008, Logemann 1998). Although scale-based subjective judgments of UES AP distension are clinically convenient and expedient during the assessment of the swallowing function in VF, objective quantification of UES opening leads to more accurate evaluation of recovery or the effects of clinical interventions designed to mitigate impaired UES function (Martin-Harris, Brodsky, Michel, Castell, Schleicher, Sandidge, Maxwell & Blair 2008, Lee, Randall, Evangelista, Kuhn & Belafsky 2017). (Omari, Ferris, Dejaeger, Tack, Vanbeckevoort & Rommel 2012) proposed to measure UES AP distension at 15-20 mm below the level of tracheal air column and calibrate the measurements by adjacent catheter sensor distances. While ASPEKT Method suggests using line between anterior-inferior corners of C2 and C4 vertebra as anatomical references,

24 the UES position is subjectively judged (C. M. Steele & Wolkin 2019). Development
25 of a validated non-invasive method that could objectively infer about swallow kinematic
26 events such as UES AP distension would add a valuable measurement tool to the dysphagia
27 diagnostic armamentarium.

28 Cervical auscultation (CA) is a well-known but crude non-invasive screening method
29 that utilizes a stethoscope to listen to swallowing sound and infer about swallow physiology.
30 Studies have provided evidence for CA's dependence on age, volume size and volume viscosity
31 (Cichero & Murdoch 2002). And dysphagic individuals were reported to have long last and
32 high-pitched swallow sounds through CA than normal subjects (Cichero & Murdoch 2006).
33 The cardiac analogy hypothesis on which CA is based germanely suggests that CA acoustic
34 signals are generated via vibrations caused by valve and pump systems within the upper
35 aerodigestive tract during different swallowing events (Cichero & Murdoch 1998, Kurosu,
36 Coyle, Perera & Sejdić 2019, Khalifa, Donohue, Coyle & Sejdić 2020). The nature of
37 this acoustic information produced by the actions of these valves has yet to be completely
38 elucidated (Sejdić et al. 2019). Although CA in clinical use is believed by devotees to
39 provide sufficient information for assessment of swallowing physiology and biomechanics, a
40 robust body of literature refutes its diagnostic value (Lagarde, Kamalski & Engel-Hoek 2016).
41 This can be referred to the narrow frequency response of the stethoscope which makes it
42 incapable of transmitting the entire spectrum of acoustic and most of vibratory information
43 emanating from the pharynx during swallowing, thus limiting the human judgment to the
44 audible differences only. CA might show fair sensitivity (ranges from 23% to 94%) and
45 specificity (ranges from 50% to 74%) in dysphagia diagnosis; however, studies have argued
46 that intra-rater reliability of CA varied widely and different raters shared poor agreement
47 while assessing the audio of the same swallows (P. Leslie & Wilson 2004, A. E. Stroud &

48 Wiles 2002, Lagarde et al. 2016).

49 High-resolution cervical auscultation (HRCA) is an equally non-invasive, promising
50 and advanced alternative to CA that employs electronic sensors (i.e., high-resolution
51 accelerometers and microphones) to transmit the entire range of acoustic and vibratory
52 information produced by the kinematic and bolus-flow events occurring during swallowing
53 (Takahashi, Groher & Michi 1994, Dudik, Jestrović, Luan, Coyle & Sejdić 2015, Movahedi,
54 Kurosu, Coyle, Perera & Sejdić 2017, Khalifa, Coyle & Sejdić 2020). Unlike CA which
55 relies solely on traditional, nonstandardized human interpretation of the swallowing sounds
56 through stethoscope, HRCA signal feature analyses are not prone to human judgment and
57 present unbiased and more reliable interpretations than is possible with conventional CA.

58 There is a growing body of literature examining and reporting the association and
59 predictive value of HRCA signal features analysis in approximating human measurements
60 for a variety of swallowing kinematic events. For instance, HRCA signal features were found
61 to be strongly correlated with human measurements of VF-based hyoid displacement during
62 swallowing (Rebrion, Zhang, Khalifa, Ramadan, Kurosu, Coyle, Perera & Sejdić 2019, He,
63 Perera, Khalifa, Zhang, Mahoney, Sabry, Donohue, Coyle & Sejdić 2019). Moreover, HRCA
64 signals have been used to non-invasively detect swallows and isolate swallow events from non-
65 swallowing activity, track the location of the hyoid bone on every frame of a VF video image
66 sequence (Mao, Zhang, Khalifa, Donohue, Coyle & Sejdić 2019), and closely approximate
67 human measurements of UES opening duration (Khalifa, Donohue, Coyle & Sejdić 2020).
68 Other studies have shown promising results in aspiration detection and categorization based
69 on the penetration-aspiration scale (PAS) (Yu, Khalifa & Sejdić 2019, Rosenbek, Robbins,
70 Roecker, Coyle & Wood 1996), unsupervised screening of healthy vs. pathological swallows
71 (Dudik, Coyle, El-Jaroudi, Mao, Sun & Sejdić 2018), and demonstrating the association

72 of HRCA signals with several swallowing events (e.g., laryngeal vestibule closure) (Kurosu
73 et al. 2019). However, to the best of our knowledge, no study has addressed the relationship
74 between UES anterior-posterior (AP) distension and HRCA signal features. In line with
75 the cardiac analogy hypothesis, we supposed that detachment of the anterior and posterior
76 walls of the UES may provide the valve activity that generates swallowing sounds and
77 vibrations which can be recorded with HRCA (Cichero & Murdoch 1998, Kurosu et al. 2019).
78 We hypothesized that HRCA signals will show strong correlation with the UES distension
79 measurements which could be revealed by statistical tests. Therefore, we sought to examine
80 whether HRCA signal features, from both acoustic and tri-axial acceleration signals, are
81 ~~statistically significant to~~ significantly associated with the VF-measured maximal AP UES
82 distension.

83 **2. Methods**

84 *2.1. Subjects, swallows, and data acquisition*

85 The study was approved by the institutional review board (IRB) of the University
86 of Pittsburgh, and all participating patients provided informed consent prior to their
87 participation. This study was performed as apart of clinical experiment conducted in the
88 context of a standard clinical swallowing evaluation procedure rather than controlled research
89 procedure. Speech language pathologists (SLPs) who conducted the experiment had full
90 control over the procedure which allowed them to alter the bolus size, consistency, maneuver,
91 and way of administration as deemed necessary and based on the patient condition. The
92 following consistencies were used during videofluoroscopy: thin liquid (Varibar thin, Bracco
93 Diagnostics, Inc., < 5 cPs viscosity), mildly thick liquid (Varibar nectar, 300 cPs viscosity),

94 puree (Varibar pudding, 5000 cPs viscosity), and Keebler Sandies Mini Simply Shortbread
 95 Cookies (Kellogg Sales Company). Boluses were either self-administered by patients via a
 96 cup or a straw or administered by the clinician through the use of a spoon (3-5 mL). Two
 97 hundred three swallows were accrued from 27 patients referred to speech-language pathology
 98 for VF evaluation of suspected dysphagia at the University of Pittsburgh Medical Center
 99 Presbyterian University Hospital (Pittsburgh, PA). Participants included 20 males [ages
 100 ranged between 41 – 86 years (mean age 64.85 ± 12.72 years)] and 7 females [ages ranged
 101 between 57 – 76 years (mean age 66 ± 7.13 years)]. Of the sample, 15 patients were diagnosed
 102 with stroke while the remaining 12 patients were diagnosed with different medical conditions
 103 unrelated to stroke. Patients’ demographics and characteristics are included in table 1.

104 VF swallowing studies were conducted while participants were sitting laterally to a
 105 standard fluoroscopic x-ray machine system (Precision 500D system, GE Healthcare, LLC,
 106 Waukesha, WI) with their head adjusted to a neutral position to have a clear view of
 107 the oral cavity, pharynx, upper esophagus and vertebral column. The VF videos were
 108 initially recorded at 30 pulses per second (PPS) and captured by a frame grabber module
 109 (AccuStream Express HD, Foresight Imaging, Chelmsford, MA) with a sampling rate of 60
 110 frames per second (FPS).

Table 1. Subject distribution.

Characteristics	Number of samples
Total subjects	27
Total swallows	203
Female subjects	7
Male subjects	20
Stroke patients	15
Non-stroke patients	12
Age range	41-86

111 HRCAs signals were obtained concurrently with VF by attaching a tri-axial accelerometer
112 (ADXL 327, Analog Devices, Norwood, Massachusetts) and a microphone (model C 411L,
113 AKG, Vienna, Austria) to the anterior neck of participants, as shown in figure 1. The
114 tri-axial accelerometer was powered at 3 volts (model 1504, BK Precision, Yorba Linda,
115 California) and attached at the midline of the anterior neck of participants over the arch of
116 the cricoid cartilage (Takahashi et al. 1994, Dudik, Coyle, Perera & Sejdić 2015). The sensor’s
117 superior-inferior (SI) or vertical axis was aligned to the vertical axis of the participant’s
118 neck in the sagittal plane, with the anterior-posterior (AP) and medial-lateral (ML) axes
119 aligned perpendicular to the SI axis in the transverse and coronal planes respectively. The
120 microphone was powered by a B29L power supply, set to volume level 9 (model B29L,
121 AKG, Vienna, Austria), and positioned over the lateral side of larynx slightly below the
122 accelerometer in a position that did not interfere with x-ray imaging of the laryngeal or
123 tracheal airway. Tri-axial accelerometry signals were then filtered by a band-pass filter from
124 0.1 to 3000 Hz with 5x amplification gain (model P55, Grass Technologies, Warwick, Rhode
125 Island). All signals extracted from the tri-axial accelerometer and microphone were fed into
126 National Instrument 6210 DAQ and recorded by Labview Program Signal Express (National
127 Instrument, Austin, Texas) at a sampling rate of 20 kHz. This setup has demonstrated
128 reliable effectiveness for swallowing activity detection in previous studies (Dudik, Kurosu,
129 Coyle & Sejdić 2016, Lee, Sejdić, Steele & Chau 2010).

130 Each participant completed a sequence of swallows, including swallows of large volumes
131 ranging from $5mL$ to $15mL$ depending on the participants’ comfort delivered by scaled
132 cup, and small volumes of $3mL$ delivered in a standard teaspoon. Barium of various
133 consistencies was provided including: thin, thick (nectar), pudding and cookie. Distribution
134 and percentage of bolus characteristics of all swallow samples are presented in table 2 .

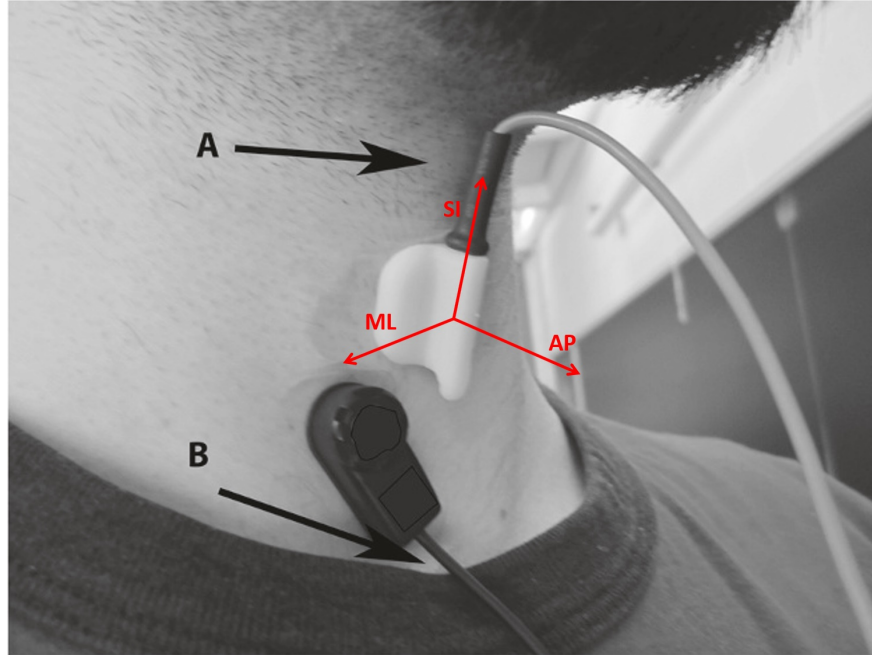
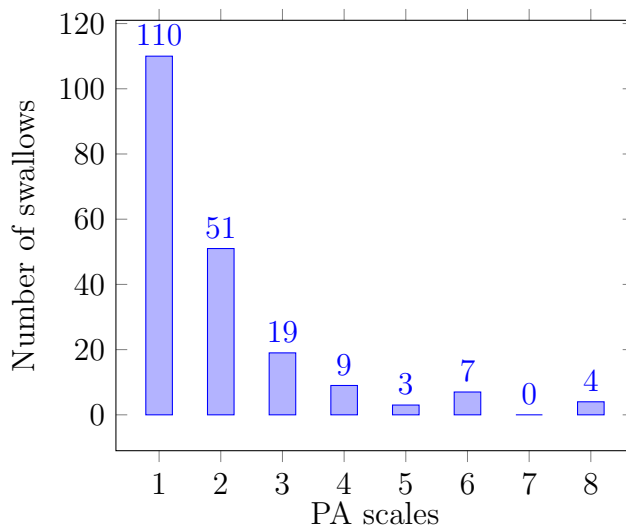


Figure 1. Positions and axes of electronic devices during the experiments. A: accelerometer attached over the arch of the cricoid cartilage on the central axis of neck; B: the microphone attached below the accelerometer and superior to the clavicle (Dudik, Jestrović, Luan, Coyle & Sejdić 2015).

135 Among the set of swallows considered in this study, airway penetration was observed in
136 7 of the 27 patients. Further, 17 patients aspirated with detailed distribution of swallows
137 according to the PAS, included in figure 2. We excluded from analysis all swallow videos
138 during which the bolus was not clearly visible on every frame during the entire swallow
139 event, and in which there was a nasogastric tube in place. The set of swallows selected for
140 this particular study varied in participant factors (e.g., age, gender, diagnosis) and bolus
141 conditions (e.g., volume, texture, mode of administration) because this research aimed to
142 examine whether HRCA signal features are associated with the AP distension of maximal
143 UES opening regardless of participant variables or characteristics of swallowed materials.

Table 2. Bolus characteristics distribution.

Bolus texture and utensil	Number of samples	Percentage of samples
Thin by cup	42	20.69%
Thin by cup with straw	17	8.37%
Thin by spoon	42	20.69%
Pudding by spoon	33	16.26%
Nectar by cup	28	13.79%
Nectar in cup with straw	25	12.32%
Cookie in spoon	9	4.43%
NA	1	0.49%

**Figure 2.** Distribution of tested swallow studies in PA scales

144 *2.2. VF image analysis*

145 All trained judges performing manual VF measurements were blinded to participant
 146 demographics, diagnosis, and bolus condition. Videos were first downsampled to 30 FPS to
 147 eliminate duplicate frames, and segmented twice into two segments of different parameters:
 148 one containing the entire swallow segment, and the other including video frames containing
 149 only the duration of UES opening. Individual swallows were segmented based on the frame
 150 in which the head of the bolus reached the ramus of the mandible (swallow onset), and the
 151 frame in which the hyoid returned to its lowest position following clearance of the bolus

152 from the UES (swallow offset) (Dudik, Jestrović, Luan, Coyle & Sejdić 2015). The other
153 UES opening segments were extracted from the frame of first UES opening (UES opening
154 onset) to the frame of first UES closure (UES opening offset). The further selection of video
155 frame containing maximal AP distension of UES along with several adjacent frames, was
156 performed as follows.

157 First, the video frame in which maximal displacement of the hyoid bone was observed
158 during the pharyngeal stage, was selected. Second, because maximal UES opening may
159 happen during, or shortly before or after maximal hyoid displacement (Jacob et al. 1989,
160 Cook et al. 1989), we measured the AP distension of UES opening in the 2-3 frames preceding
161 maximal hyoid displacement, the frame with maximal hyoid displacement, and the 2-3 frames
162 following maximal hyoid displacement (this was about 5-7 frames). The AP distension of the
163 maximal UES opening in each video was measured using an application (UES AP distension
164 drawing application) that was developed in Matlab to execute this function through the
165 following steps (figure 3):

- 166 (i) Each video was uploaded into the UES AP distension drawing application.
- 167 (ii) The frames used for measuring the AP distension of UES opening were selected. The
168 AP distension of UES opening was measured using a drawing tool for 2 or 3 frames
169 before and after maximal hyoid displacement, as well as on the frame with maximal
170 hyoid displacement.
- 171 (iii) In order to standardize judgments regarding the location of the superior and inferior
172 limits of the height of the UES, we used the height of the third cervical vertebral
173 body. The region of the proximal esophagus considered the UES has been quantified in
174 manometric studies as coursing 1.3 cm inferiorly from the base of the plane of the true

175 vocal folds (Cook et al. 1989). The height of the third cervical vertebra ranges from
176 1.11 – 1.14 cm in adult females and 1.24 – 1.37 cm in adult males based on midsagittal
177 x-ray measurements (Katz, Reynolds, Foust & Baum 1975). Therefore for each selected
178 frame, a yellow line was drawn from the anterior superior edge to the anterior inferior
179 edge of third cervical vertebral body (C3) in figure 3a.

180 (iv) Next, another red line was drawn between the anterior inferior edge of the second cervical
181 vertebrae (C2) and the anterior inferior edge of the fourth cervical vertebrae (C4) to
182 provide a vertical axis (C2-C4) that enables the algorithms to subtract larger scale
183 head/neck movements from the measurements (Molfenter & Steele 2014)(figure 3b).
184 The length of the C2-C4 segment was also used as an anatomical scalar representing
185 each subject’s height.

186 (v) The yellow line drawn in step iii was dragged and anchored to the superior border of the
187 posterior tracheal air column (indicating the range of UES height to limit judgments of
188 UES opening) as shown in figure 3c.

189 (vi) A long blue line that is perpendicular to the C2-C4 segment was drawn and used as a
190 referent axis to ensure alignment of the vertical and horizontal axes of measurement to
191 participant position rather than to an arbitrary x-y coordinate system based on strict
192 vertical and horizontal geometric axes of zero and 90 degrees. This line was then dragged
193 superiorly and inferiorly between two ends of the dragged C3 segment to the location
194 of maximal anterior-posterior distance of the UES opening (figure 3d).

195 (vii) Then, the anterior and posterior points of UES opening on the perpendicular line were
196 marked by short blue line segments respectively on the line segment drawn in step 6
197 (represented by two short parallel blue lines in (figure 3e).

198 (viii) The coordinates of the measured length of maximal UES opening (UES opening anterior
199 end X and Y, UES opening posterior end X and Y) were returned in the output of the
200 application.

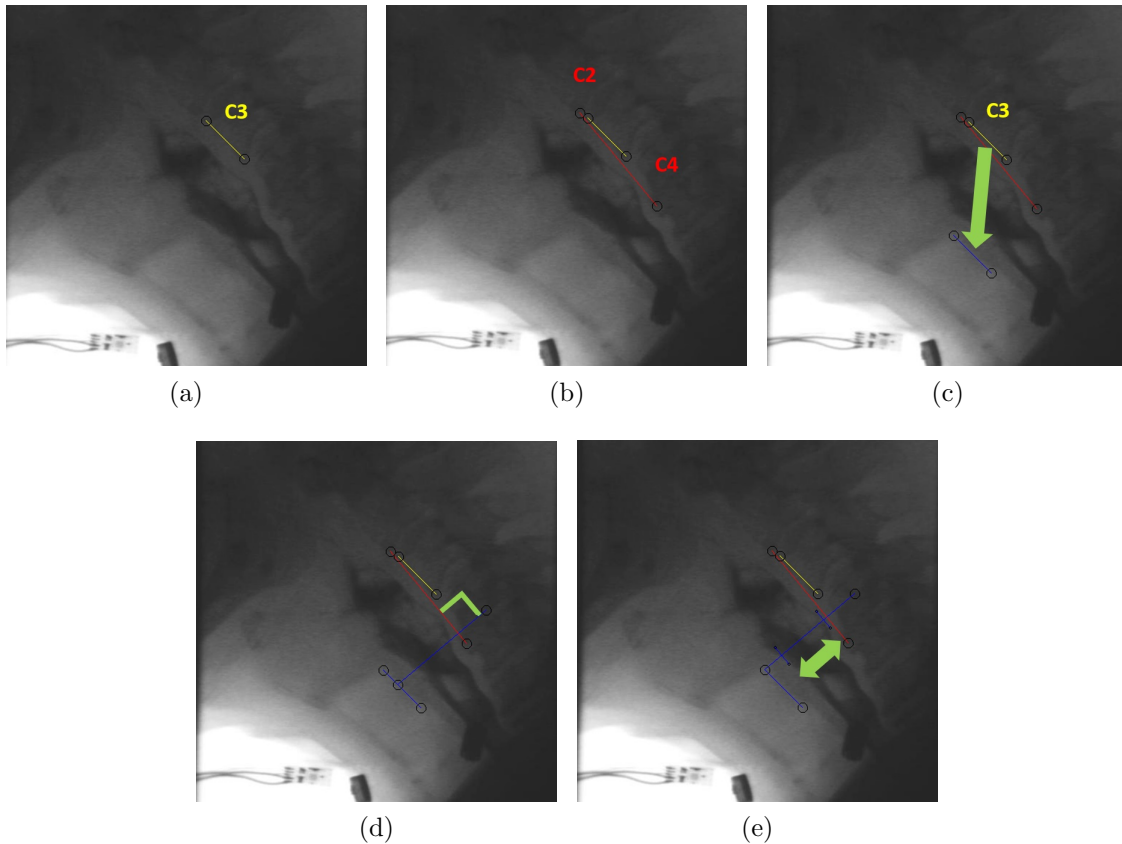


Figure 3. Illustration of steps for measuring the AP distension of maximal UES opening using the newly developed UES AP distension drawing application: (3a) The length of anterior edge of C3 is indicated by the yellow line segment; (3b) The anterior inferior edge of C2 and the anterior inferior edge of C4 were connected by the red line segment; (3c) The C3 length was dragged, following the green arrow, to the position of blue line segment with the upper ends anchored to the superior border of tracheal air column; (3d) The longer blue line segment perpendicular to the C2-C4 axis was positioned with its left ends sliding on the dragged C3 segment; (3e) When the reference line (longer blue line segment) was adjusted to across the largest width of UES opening, two short blue line segments were placed on the extremities of UES. The length of UES opening is measured between the two short segments represented by the bidirectional green arrow.

201 Interclass correlation coefficients (ICC) have been ubiquitous in assessing the reliability
202 of human judgements for VF analysis in dysphagia research. In this study, the two judges'

203 level of agreement of the manual measurements described above underwent testing of intra-
204 and inter-judge reliability from 10% of the measured swallows using the absolute agreement
205 ICC on an ongoing basis during measurements of study data, to mitigate judgment drift
206 (Shrout & Fleiss 1979). Excellent ($ICC > 0.98$) intra- and inter-judge reliability between
207 the judges of all VF analysis including temporal (i.e., video segmentation, identification
208 of first UES opening and first UES closure, identification of maximal hyoid displacement),
209 as well as spatial (i.e., UES AP distension measurements). The difference between UES
210 distension measures during the inter-rater reliability test was illustrated by a Bland-Altman
211 plot as shown in Figure 4. Consistency of proposed UES distension measuring method was
212 demonstrated and the UES distension measurements were validated for further statistical
213 analysis.

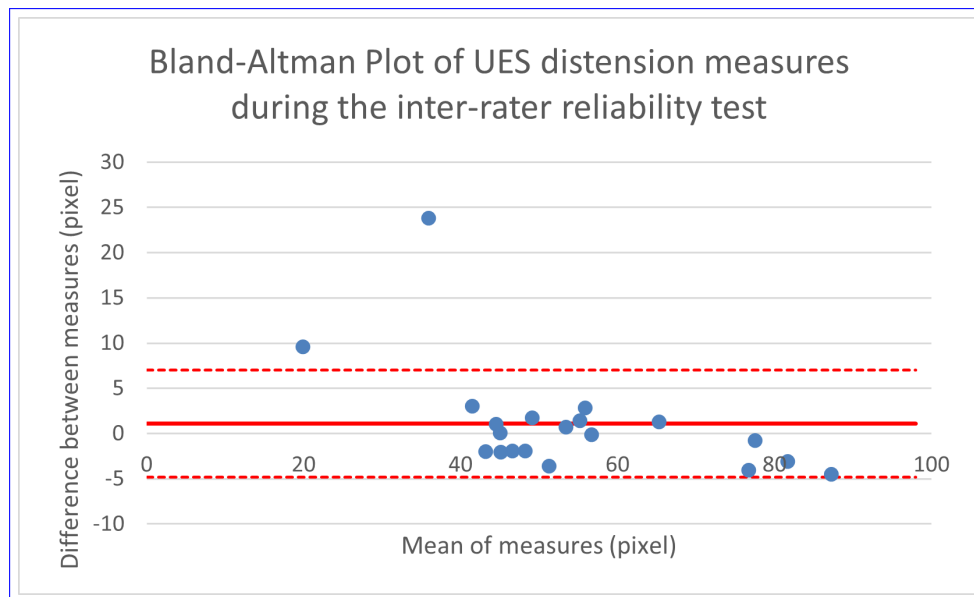


Figure 4. Bland-Altman plot of UES distension measures: the red solid line indicates the average difference between measures and red dotted lines correspond to .95 confidence interval on the average

214 2.3. Signals preprocessing and feature extraction

215 Independent of the temporal and spatial measurements described above, processing and
216 extraction of signal features was performed. The preprocessing of HRCA signals consists of
217 several steps: first, all the signals, recorded at a sampling rate of 20 kHz, were downsampled
218 to 4 kHz to overcome the disturbing noise in the signals due to multiple environmental sources
219 and other measurement errors. Additionally, to improve the quality of the collected signals,
220 the device noise inherent with both the accelerometer and the microphone, was modeled
221 by fitting an autoregressive model to the sensor signals generated with null input. Model
222 coefficients were used to generate sensor and axis-specific finite impulse response filters to
223 reduce the device noise (Sejdić, Kosmisar, Steele & Chau 2010). Then, the low-frequency
224 components and motion artifacts were eliminated from accelerometer signals using fourth-
225 order least-square splines (Sejdić, Steele & Chau 2012, Sejdić, Steele & Chau 2010a). Lastly,
226 the effect of broadband noise on signals was reduced using the tenth order Meyer wavelet
227 decomposition (Sejdić, Steele & Chau 2010b).

228 The cleaned swallowing signals, including the swallowing acoustic signal (obtained
229 through the microphone) and the acceleration signals of 3 accelerometer axes [anterior-
230 posterior (AP), superior-inferior (SI), and medial-lateral (ML)], passed through the feature
231 extraction stage. The features extracted from time, frequency and time-frequency domains
232 were later compared to the AP distension of maximal UES opening, measured from the
233 spatial analyses using the UES AP distension drawing application. We determined which
234 signal features are related to the AP distension of maximal UES opening and how they are
235 related.

236 Regarding time-domain features, we calculated the standard deviation, skewness, and

237 kurtosis which evaluate the extent of deviation, asymmetry, and sharpness of the peak,
 238 of the statistical distribution of signals respectively. We then determined the Lempel-Ziv
 239 complexity by using Kolmogorov complexity which measures the regularity of the signal
 240 and the entropy rate which represents the amount of information contained by the signal.
 241 Furthermore, in the frequency domain, we took the peak frequency, spectral centroid, and
 242 bandwidth into consideration. While the peak frequency and spectral centroid indicate the
 243 spectral performances of the signal, the bandwidth specifies the distribution of the power
 244 spectrum relative to the spectral centroid. Finally, we estimated the entropy rate of a tenth
 245 order discrete Meyer wavelet decomposition to provide information about the distortion of
 246 signals in the time-frequency domain. The efficiency and a detailed description of the above
 247 features (summarized in table 3) was presented in previous work (Rebrion et al. 2019, He
 248 et al. 2019, Sejdić, Kosmisar, Steele & Chau 2010, Sejdić, Steele & Chau 2010a).

Table 3. Definitions of extracted features.

	Feature	Definition
Time domain	Standard deviation	Variation of the signal around mean value
	Skewness	Asymetry of statistical distribution of the signal
	Kurtosis	Sharpness of the peak of signal amplitude distribution
Information-theoretic domain	Lempel-Ziv complexity	Regularity of the signal
	Entropy rate	Randomness of the signal
Frequency domain	peak frequency	Frequency that corresponds to the maximal spectral energy
	Centroid frequency	Frequency that divides the spectrum into two equal parts
	Band width	Difference between the uppermost and lowermost frequencies of the signal spectrum
Time-frequency domain	Wavelet entropy	Disordered/ordered behavior of the signal

249 Finally, to test whether there were differences in the strength of associations between
 250 signal features and UES opening AP distension when analyses were conducted using the
 251 entire swallow event segment versus solely the shorter sub-duration of the UES opening
 252 event, the previously described features were extracted from signals recorded from the whole
 253 swallowing segment as well as from signals recorded only during UES opening segment.

254 *2.4. Statistical analysis*

255 To examine the association between HRCA signal features and the AP distension of maximal
256 UES opening, we fitted a series of linear mixed models with the UES AP distension as the
257 response variable; each of the HRCA signal features, one at a time, as independent variables;
258 and a participant random effect to account for multiple swallows from the same individual.
259 To adjust statistical analysis for multiple testing, false discover rate (FDR) was applied to
260 each testing result (Benjamini & Hochberg 1995). We used SAS[®] version 9.3 (SAS Institute,
261 Inc., Cary, North Carolina) for analysis.

262 **3. Results**

263 *3.1. Signal features analysis*

264 *3.1.1. Time-domain signal variability* Tables 4 and 5 show the raw means and standard
265 deviations of signal features extracted from the microphone and accelerometer (swallowing
266 sound, AP, SI and ML vibrations), by using whole swallow segment data and UES opening
267 segment data respectively. In both cases, all signal features were found to have low standard
268 deviation, low skewness (i.e., contain both positive and negative values) and high kurtosis,
269 indicating that the signals were distributed evenly around the mean value and exhibited
270 peak variation at certain times. The SI movement had a higher standard deviation than AP
271 and ML directions which implies that the dominant vibrations were caused by the superior
272 hyolaryngeal displacement rather than other movements. This finding is consistent with
273 previous HRCA studies on hyoid excursion (Movahedi et al. 2017, Rebrion et al. 2019, He
274 et al. 2019, Zoratto, Chau & Steele 2010). Signals from whole swallow segment data and
275 UES opening segment data had close values on standard deviation and skewness but differed

276 in kurtosis. UES opening segment data showed less kurtosis thus a smoother sequence in
277 the time-domain.

278 *3.1.2. Information theory based features* Both accelerometry and sound signals showed low
279 Lempel-Ziv complexity and high entropy rate which corresponds to a rather low randomness
280 and high predictability. However, among the vibration signals, the SI signal presented
281 the greatest Lempel-Ziv complexity, and therefore less predictable behavior than ML and
282 AP signals. Meanwhile, the sideways vibration (i.e., ML signals), showed greater entropy
283 rate and were more regular than upward SI and forward movement (i.e., AP) signals.
284 Compared to the tri-axial accelerometry, swallow sound signals had the most complexity
285 and least regularity. Furthermore, UES opening segment signals showed greater Lempel-Ziv
286 complexity and less entropy rate than the whole swallow segment signals.

287 *3.1.3. Frequency domain distribution* When considering the distribution in the frequency
288 domain, all signals presented low peak frequency, high bandwidth, and high centroid
289 frequency indicating that the signals had similar energy distribution over the frequency.
290 While peak frequencies varied to a small extent, both the bandwidth and centroid frequency
291 of HRCA signals decreased when narrowing down the data from UES opening onset to
292 UES closure. In particular, the bandwidth of the ML signal changed tremendously. Both
293 AP and ML signals were more widely spread over the frequency domain than SI signals
294 and swallow sounds considering the whole swallow segment. Moreover, AP signals had the
295 largest distribution over frequency and exhibit greater centroid frequency than others in both
296 segments. Lastly, in the time-frequency domain, the reduction of wavelet entropy indicates
297 that the data extracted from the UES opening segment led to more orderly behavior than
298 the whole swallow segment in all HRCA signals.

Table 4. Mean and standard deviations of considered features from whole swallowing data (SW). (Mic) represents the recorded swallow sound; (AP) represents the accelerometer variation over Anterior-posterior axis; (SI) ~~corresponds to~~ stands for the acceleration over Superior-interior axis and (ML) ~~stands for~~ corresponds to Medial-lateral direction.

	MIC_{sw}	AP_{sw}	SI_{sw}	ML_{sw}
Time domain				
Standard deviation	0.014 ± 0.009	0.021 ± 0.019	0.042 ± 0.024	0.012 ± 0.008
Skewness	-0.433 ± 2.917	0.491 ± 3.868	-0.456 ± 1.992	-0.056 ± 2.432
Kurtosis	37.900 ± 89.202	54.812 ± 134.569	20.432 ± 51.416	29.705 ± 83.436
Lempel-Ziv complexity	0.240 ± 0.083	0.161 ± 0.068	0.216 ± 0.067	0.173 ± 0.068
Entropy rate	0.912 ± 0.041	0.945 ± 0.026	0.939 ± 0.023	0.952 ± 0.022
Frequency domain				
Peak frequency (Hz)	18.145 ± 28.090	10.363 ± 21.343	8.425 ± 8.692	9.357 ± 10.462
Bandwidth (Hz)	108.765 ± 28.090	167.814 ± 152.323	71.258 ± 76.123	156.279 ± 145.890
Centroid frequency (Hz)	84.494 ± 64.215	121.917 ± 170.669	42.471 ± 68.798	69.908 ± 102.359
Time-frequency domain				
Wavelet entropy	1.171 ± 0.816	0.904 ± 0.744	0.880 ± 0.744	0.785 ± 0.733

Table 5. Mean and standard deviation of same features obtained from UES opening (UESO) data segment.

	MIC_{ueso}	AP_{ueso}	SI_{ueso}	ML_{ueso}
Time domain				
Standard deviation	0.013 ± 0.010	0.019 ± 0.018	0.043 ± 0.027	0.011 ± 0.007
Skewness	-0.430 ± 2.213	0.556 ± 2.567	-0.394 ± 1.540	0.034 ± 1.602
Kurtosis	23.821 ± 47.005	31.466 ± 60.382	12.348 ± 21.019	13.891 ± 37.620
Lempel-Ziv complexity	0.287 ± 0.093	0.213 ± 0.082	0.264 ± 0.074	0.243 ± 0.071
Entropy rate	0.878 ± 0.060	0.915 ± 0.039	0.909 ± 0.034	0.919 ± 0.032
Frequency domain				
Peak frequency (Hz)	19.717 ± 28.959	14.113 ± 44.726	9.447 ± 12.904	8.944 ± 10.230
Bandwidth (Hz)	94.508 ± 81.338	121.633 ± 150.981	57.583 ± 63.966	59.911 ± 79.896
Centroid frequency (Hz)	76.397 ± 72.924	110.857 ± 198.028	36.924 ± 55.771	36.535 ± 59.638
Time-frequency domain				
Wavelet entropy	0.973 ± 0.746	0.737 ± 0.671	0.708 ± 0.603	0.716 ± 0.665

299 3.2. Maximal UES AP distension is associated with HRCA signal features

300 3.2.1. Maximal UES AP distension measurements Table 6 depicts the difference in the

301 values of mean and standard deviation between the original AP distension of maximal UES

302 opening in pixels, the values normalized to the length of C2-C4, and the values normalized
 303 to the anterior length of C3. Interestingly, the maximum AP distension of UES opening
 304 closely approximated the height of the anterior plane of the C3 vertebral body.

Table 6. Mean and standard deviation of maximum anterior-posterior AP distension of UES.

	Values
Maximum AP distension width (Pixel)	50.935 ± 16.513
Maximum AP distension width normalized to one C2-C4 length	0.362 ± 0.117
Maximum AP distension width normalized to one C3 length	0.937 ± 0.312

305 *3.2.2. Statistical significance between UES distension and HRCA signal features*

306 Following statistical analysis, significant correlations were demonstrated between certain
 307 features extracted from the whole swallow signals and the values of the AP distension of
 308 maximal UES opening divided by C2-C4 length, as depicted in table 7. These features, with
 309 a FDR adjusted p-value $< .05$, include the following; standard deviation and Lempel-Ziv
 310 complexity of swallow sound; standard deviation and wavelet entropy of AP signal; standard
 311 deviation, skewness, Lempel-Ziv complexity and peak frequency of SI signal, and finally
 312 standard deviation of ML signal.

313 When comparing associations between signal features and maximum UES opening
 314 based on the entire swallow event segment versus solely the duration of UES opening,
 315 more HRCA features were found to be significantly correlated to the AP distension of the
 316 maximal UES opening divided by C2-C4 length from UES opening segment, as shown in
 317 table 8. Lempel-Ziv complexity and entropy rate of all signals (sound, AP, SI, and ML)
 318 were significantly correlated to the measured UES AP distension. Additionally, skewness of
 319 swallow sound, bandwidth of AP vibrations, kurtosis and centroid frequency of SI vibrations

Table 7. Statistical significance and regressive coefficients between the maximum UES AP distension [normalized to C2-C4 length \(Dmax/C2C4\)](#) with the accelerometer and the microphone signal features using whole swallow data (NS stands for not significant).

	Dmax/C2C4			
	MIC_{sw}	AP_{sw}	SI_{sw}	ML_{sw}
Standard deviation	$p = 0.0185$ $Coef = 2.2971$	$p = 0.0005$ $Coef = 1.4111$	$p = 0.0195$ $Coef = 0.8187$	$p = 0.0272$ $Coef = 2.7806$
Skewness	NS -	NS -	$p = 0.0195$ $Coef = -0.0009$	NS -
Kurtosis	NS -	NS -	NS -	NS -
Lempel-Ziv complexity	$p = 0.0434$ $Coef = -0.1980$	NS -	$p = 0.0432$ $Coef = -0.2345$	NS -
Entropy rate	NS -	NS -	NS -	NS -
Centroid frequency	NS -	NS -	NS -	NS -
Peak frequency	NS -	NS -	$p = 0.0432$ $Coef = -0.0018$	NS -
Bandwidth	NS -	NS -	NS -	NS -
Wavelet entropy	NS -	$p = 0.0432$ $Coef = 0.0201$	NS -	NS -

320 became significant to the normalized AP distension of maximal UES opening, while skewness
321 and peak frequency of SI vibrations lost their importance. Far more features were found to
322 be related to the UES opening segment data.

323 3.2.3. Linear regressive coefficients between UES distension and HRCA signal features

324 Regressive coefficients (noted as Coef) in Tables 7 and 8 refer to how the maximal UES
325 AP distension (normalized to C2-C4 length) varies across the HRCA signal features, as
326 suggested by the linear mixed model established previously. For instance, if the standard
327 deviation of the sound signal in a whole swallowing segment is increased by 1 unit, then

Table 8. Statistical significance and regressive coefficients between the maximum UES AP distension [normalized to C2-C4 length \(Dmax/C2C4\)](#) with the accelerometer and the microphone signal features using UES opening data (NS stands for not significant).

	Dmax/C2C4			
	MIC_{ueso}	AP_{ueso}	SI_{ueso}	ML_{ueso}
Standard deviation	$p = 0.0100$ $Coef = 2.2571$	$p = 0.0069$ $Coef = 1.4044$	$p = 0.0100$ $Coef = 0.7907$	$p = 0.0185$ $Coef = 3.1927$
Skewness	$p = 0.0352$ $Coef = -0.0079$	NS -	NS -	NS -
Kurtosis	NS -	NS -	$p = 0.0100$ $Coef = 0.0010$	NS -
Lempel-Ziv complexity	$p = 0.0020$ $Coef = -0.2956$	$p = 0.0048$ $Coef = -0.2969$	$p = 0.0051$ $Coef = -0.3096$	$p = 0.0100$ $Coef = -0.3059$
Entropy rate	$p = 0.0100$ $Coef = 0.3561$	$p = 0.0288$ $Coef = 0.4549$	$p = 0.0204$ $Coef = 0.5520$	$p = 0.0185$ $Coef = 0.5889$
Centroid frequency	NS -	NS -	$p = 0.0185$ $Coef = 0.0004$	NS -
Peak frequency	NS -	NS -	NS -	NS -
Bandwidth	NS -	$p = 0.0185$ $Coef = 0.0002$	NS -	NS -
Wavelet entropy	NS -	$p = 0.0434$ $Coef = 0.0220$	NS -	NS -

328 the UES AP distension divided by C2-C4 length will increase by 2.2971 units. Negative
329 values mean that the UES AP distension will decrease if the value of the considered feature
330 increases.

331 4. Discussion

332 4.1. Objective measurement of UES AP distension

333 In this study, we presented a standardized method of maximal UES anterior-posterior
334 AP distension measurement in lateral view VF which calibrates against subject height
335 and corrects for any possible head or neck movement by using C2-C4 normalization and

336 supplement referent axis to indicate subject's vertical axis. The resulting measurement
337 of UES AP distension was placed in proximity to anterior planes of C5 and C6 which is
338 coherent to the protocol of UES opening measurement in Omari et al.'s (2012) and Kim
339 et al.'s (2015) studies. The ratios between values of the AP distension of maximal UES
340 opening and C3 length are close to 1, which indicates that the C3 length may be a potential
341 anatomic reference of sufficient UES opening that enables clinicians objective evaluate UES
342 opening during the VF study thus leads to trial interventions. We plan to investigate the
343 feasibility and reliability of such a reference in VF analysis of images obtained from healthy
344 participants in the future.

345 *4.2. Features extracted from HRCA signals*

346 Both mean and standard deviation for most of the features extracted from HRCA
347 recordings exhibit consistency with the previous results (Movahedi et al. 2017, Rebrion
348 et al. 2019, Dudik, Coyle, Perera & Sejdić 2015) in terms of magnitude. On the other
349 hand, the UES opening segments of HRCA signals present smoother distribution in the
350 time domain, narrower distribution in the frequency domain, as well as, greater complexity
351 and randomness than the whole swallow segments. This observation reflects that complex
352 swallowing events occur during UES opening onset.

353 *4.3. AP distension of maximal UES opening and HRCA signal features*

354 A robust portion of HRCA signal features extracted from whole swallow segments was
355 significantly correlated to the normalized AP distension of maximal UES opening. However
356 more signal features were significantly relevant to the maximal UES AP distension when
357 we performed the analysis from the shorter-segments of UES opening duration alone. This

358 finding is reasonable since the UES AP distension equals zero outside the UES opening
359 segment. Still, additional work to ascertain whether UES opening onset segmentation
360 contributes to more accurate UES distension prediction is warranted as this would represent
361 an important and clinically useful automated measurement adjunct to VF interpretations.
362 Several features from AP vibrations turned out to be ~~statistically significant to~~ significantly
363 associated with the UES AP distension, which is expected given the fact that the UES
364 distends in the AP direction to allow the passage of bolus during swallowing (Jacob
365 et al. 1989, Kahrilas 1997). The swallowing sound, generated predominantly by the valve
366 and pump activity caused by UES opening and closure, also contained more influential
367 HRCA features from UES opening segments rather than whole swallow segments (Cichero
368 & Murdoch 1998).

369 In aspects of specific features, the AP distension of maximal UES opening was more
370 associated with Lempel-ziv complexity and entropy rate in all HRCA signals, indicating that
371 the amount of information and the extent of complexity of UES opening segments strongly
372 reflects the evaluation of UES AP distension. Our statistical analysis demonstrated that a
373 significant association between HRCA signals and the AP distension of maximal UES opening
374 exists. In addition, narrowing down the segment under investigation from HRCA signals to
375 UES opening only helps strengthen the association by introducing a larger proportion of
376 signal features that were influential to the estimation of UES AP distension.

377 **5. Conclusion**

378 This study provides evidence that a significant number of HRCA signal features are
379 statistically correlated to the maximal UES AP distension measured from VF. This
380 preliminary finding demonstrates the feasibility of potential HRCA signals analysis

381 algorithms in the non-invasive prediction of maximal UES AP distension during swallowing.
382 Further research may investigate the interpretation of detailed swallow events from HRCA
383 signals and developing HRCA screening into a more effective and accurate dysphagia
384 assessment tool in clinical practice.

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