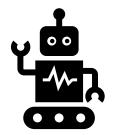
2nd UEP/EAP-BLA joint meeting

Zagreb, May 23-25th 2024

Voice related biomarkers

Mieke Moerman







2nd UEP/EAP-BLA joint meeting

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Voice related biomarkers

the UEP biomarkercommittee:

Abdelgoad Ahmed Nashaat Neveen Valentina Camesasca Daniele Farneti hernandez Villoria Ramon Ilona Kaminska Gavkhar Khaydarova

Mieke Moerman Haldun Oguz Paderno alberto Mette Pedersen Virgilijus Uloza Necati Enver Das Sneha





Biomarker

In 1998, the National Institutes of Health Biomarkers Definitions Working Group defined a biomarker as "a characteristic that is objectively measured and evaluated as an indicator of normal biological processes, pathogenic processes, or pharmacologic responses to a therapeutic intervention."

Molecular, histologic, radiographic, or physiologic characteristics are types of biomarkers.

prognostic/predictive value & monitoring value





1. Specificity





1. Specificity

2. Effectiveness









1. Specificity

2. Effectiveness

3. Efficiency











Author(s)	Year	Title	Abstract
Idrisoglu et al.			This systematic literature review (SLR) investigated the state of the art of voice-based diagnostic and n
Worasawate et al.		2023 Classification of Parkinson's Disease from Smartphone Recording Data Using Time-frequency Analysis and Convolutional Neural Network	A total of 29,798 ten-second voice recordings on smartphone from 4,051 participants were used for the
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Ngo et al		2022 Computerized analysis of speech and voice for Parkinson's disease: A systematic review	
Ma et al		2021 Deep dual-side learning ensemble model for Parkinson speech recognition	
Sahandi et al		2021 Exploring Test-Retest Reliability and Longitudinal Stability of Digital Biomarkers for Parkinson Disease in the m-Power Data Set: Cohort Study	Among the features differing between PD and HC, only a few tapping and voice features had good to e
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UEP BMC _WP1: literature _ Ramon Hernandez Villoria; Mette Pedersen, Neveen Nashaat



Phoniatrics

The medical discipline regarding communication (voice, speech, language & hearing) and swallowing problems





Glottal Function

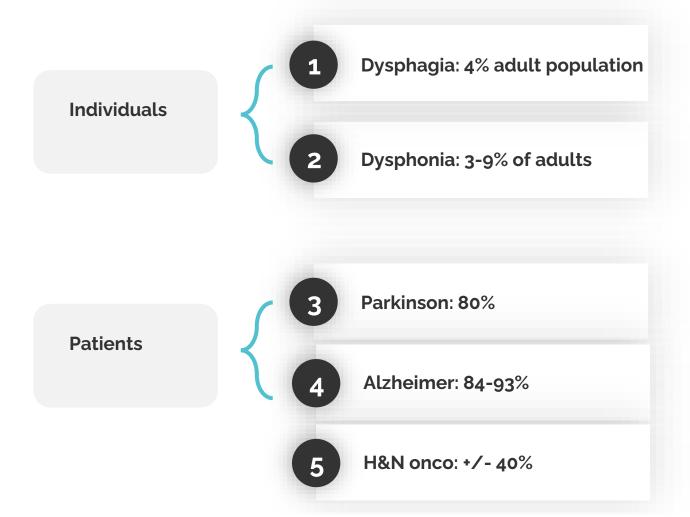
Voice & swallowing



The three most important and highly relevant functions of the glottis are to facilitate **ventilation**, facilitate **phonation**, and provide **airway protection**.



Prevalence of swallowing & voice problems



Prevalence of swallowing & voice problems

Dysphagia: 4% adult population **Individuals** Dysphonia: 3-9% of adults Parkinson: 80% **Patients** Alzheimer: 84-93% H&N onco: +/- 40%

Country	Population	# Adults (25-65j)	Dysphagia (4% of adults)	Dysphonia (5%*37,5%)
Belgium	11,6M	52%	240K	217K
The Neth	17,5M	52%	364K	328K
Germany	83M	53%	1,80M	1,5M
US	332M	65%	8,6M	6,2M

Region	Population	Parkinson	Alzheimer	H&N onco
Europe	746M	1,2M	9,7M	450K
US	332M	1M	6,2M	66K
Region	Population	Parkinson (80%)	Alzheimer (90%)	H&N onco (40%)
Region Europe	Population 746M			

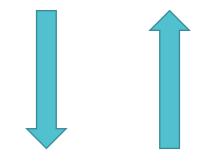
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4			





Voice related biomarkers

Disease



Voice Change





Voice related biomarkers

Disease



- Up to 78% of early stage **Parkinson-patients**
- **NDD:** early diagnostic





47 papers with Voice Parameters in Parkinson's Disease from 2013 to 2019, 3 including AI

Author	Number Year Pa	atient nu prospective	Randomize Ca	se/Control Re	etrospective	HNR SN	IR FO	(st) I	ntensity	МРТ	IITTER APS/	% SHIMMER APS	% Snekt LTAS	CEPSTRUM	VRP	Telephone	Praat	VHI	GRBAS	Deep Brain.s	ΔΙ	Deep Learning	laryngoscopic	Software	Others
Louis, E. et al	98 2013	85	Randonnize Ca	100	1	IIIVIN SIV	ik 10((30)	itensity	IVIF	JIII LIK AF 3/	O STINVINIER AFS	70 Spekt LIA	CEFSTROW	VICE	relephone	riaat	VIII	1	Deep Brain.s		Deep Learning	iai yiigoscopic	Joitware	Others
Bauer, V. et al	97 2013	22	nlı	us cc	1														1						
Péron, J. et al	96 2013	22		us cc					1												,				
	95 2013	7	pic	us cc		1		1	1			1	1	1				1			-				
Bang, Y. et al	94 2013	60		48		1		1	1			1	1	1				1							
Teixeira, E. et al								1	1																
Silbergleit, AK. et al	93 2014	27		22				1				1	1			1									
Jafari, A. et al	92	25		10		1		1							1										1
Smith, L. et al	91 2014	28		10										1											
Yang, S. et al	90 2014	26		22							1	4										1 Kernal/SVM			Consort Warrefrom Archinis Bo
Silbergleit, AK. et al	89 2015	26		22			1				1	1				1									Cspeech Waweform Analysis Pr
Soares, DP. et al Spazzapan, EA. et al	88 2015 87 2015	22 19 1	1		1	1		1	1			1	1					1				1 peak-to peak amp var			
						-		- 1	1			1						1				T peak-to peak amp var		1	CO dns/40 Mad
Tanaka, Y. et al	86 2015	108		4.5															1					1	68 dps/40 Med
Manor, Y. et al	85 2015	21 1	1	11				1	1	1 1	1	1	1					1							22 des /25 Mard
Tsuboi, T. et al	84 2015	47																		1		1	1	1	22 dps/25 Med
Crino, C. et al	83 2016	_									1								1						Acoust analysis
Watts, C. et al	82 2016	78			1				1																
Postuma, R.B	81 2016																								Editorial
Gillivan-Murphy, P.	80 2016	30 1	1	28															1					1	
Abrahao, L.J. et al	79 2016	15 1	1	5				1										1							Pharynx Pressure
annito, M.P. et al	78 2016	16			1	1								1											H1, H2, F3
ernier, L.S. et al	77 2016																								Reaview
Neves, MRL. et al	76 2016	46				1		1	1	L								1		1					
Novotný, M. et al	75 2016	37 1	1	37																					1/3-octave band
Majdinasab, F. et al	74 2016	27 1	1	21		1		1				1	1					1							
Roubeau, B. et al	73 2016																	:	1					1	Acoustanal
Sidits, D. et al	72 2017						1	1				1	1							1	1				
Wu, Y. et al	71 2017							1	1	ı												1		SVM	
Stegenmöller, E.L. et a	70 2017																								VoicequI
Parveen, S. et al	69 2017																	:	1						
Butala, A. et al	68 2017	30	1	32		1			1	L		1	1												singing, crossovers
Da Silva, V.G. et al	67 2017	10 1	1			1		1	1	ı		1	1												Tube Treatment
kacha, A. et al	66 2017	205		74		1						1													No statistical difference
Lechien, J.R. et al	65 2018	20 1	1	10		1					1	1						:	1						Early diagnosis
Abur, D. et al	64 2018	16		19				1																	JND paradim, feedback
/ieira, M. et al	63 2018	23 1	1	20				1																	Vawel lenghtining
Motto, S. et al	62 2018	15 1	1	15					1		1								1						MESGP,MPR
Lechien, J.R. et al	61 2018																							1	Review, 1980-2017
Abur, D. et al	60 2018	20 1	1	23			1																		loudness slopes
o ,E.j. et al	59 2018	30									1														plus swallowing
lan, E.y. et al	58 2018	30									1							1 :	1						Singing
Manor, Y. et al	57 2018	26 1	1	13															1	1					- signing
		20		13				1											•	1					review 1960-2016, meta A
Pinho, P. et al	56 2018	20		20				1	1															-	review 1960-2016, meta A
Gillivan-Murphy, P.	55 2019	38 1	1	28															Ţ						
hen, J. et al	54 2019	52 1	1	32		1		1				1	1	1					1						formant ratio
Saffarian, A. et al	53 2019	23 1	1 1																1						treatment
Romann, A.J. et al	52 2019	16						1												1	1				

Mette Pedersen MD PhD and Vitus Girelli Meiner IT-University of Copenhagen

Author	Number	Year Pa	atient nu	prospective	Randomiz	ze Case/Contro	ol Retrospective	HNR S	SNR FO	O(st) Inter	nsity MPT	JITTER APS/%	SHIMMER APS/%	Spekt LTAS	CEPSTRUM	VRP	Telephone	Praat V	HI GRBAS	S Deep Brain.s	Al Deep Learning	laryngoscopic	Software	Others
	_																							3+A51:AB10207 measures,
Arora, S. et al		2019	148		1													1			1		Random forest	crossover
Behroozmand, R. et al		0 2019	1	U		1				1			1								1			
inger, M.E. et al		9 2019	2																1					
(arlsson, F. et al		18 2019	2		1			1				1	1	1					1					Dhanation quationt
echien, J.R. et alt nanor, Y. et al		7 2019 6 2019	10-				82			1		1	1	1					1	1	1			Phonation quotient
Sheiban, R. et al		5 2019	10	•			02			1		1							1	1	1		Class label prediction	Poot Moon Score
Tamplin, J. et al		14 2020	7.	5			44			1											1		Class label prediction	Singing, VQOL
Viswanathan, R. et al		3 2020	2				22		1	1											1		SVM	Glottal Closure qu
nakayama, K.		2020	2				22		-	1	1										1		SVIVI	Treatment
Ma, A. et al		1 2020	303							-	-													Review
Morello; A.N.D.C		0 2020	303	_																1	1 1		acoustanal	acoustanal
Chiaramonte, R. et al		9 2020																		•			acoustanti	Review, 14 pub, meta
Viswanathan, R. et al		88 2020	2	6			22														1		LASSO ranking	neview, 14 pab, meta
Altay, E.V. et al		7 2020	_	_																	1		nicgar, Voice data	
Park, J.E. et al		6 2020	4	7											1						1		megar, voice data	Comparred to 39 tremor
Sarac, E.T. et al		35 2020	1					1		1			1	1				1						22paca to 55 ticinol
Reyes, A. et al		34 2020	3			1		1		-		1		-				1					peak subg press	Treatment
Lechien, J.R. et al		3 2020	3:			1	20	1		1	1	-											peak subg press	reatment
Gaballah, A. et al		2021	5				51	1								1							SVR/RPDE	
Lechien, J.R. et al		1 2021	J		1		51	- 1															511y111 52	Acoust meassurements
lain, A. et al		30 2021	1	4																	1	1	p-CRNN	Ative phon plus pros feature
Gaballah, A. et al		9 2021	5						1						1	1					1	-	svr	VAT, RPDE
Rajasekar, S.S		8 2021	2						1	1	1		1	1							1			AdaBoost classifier
Da Silva,j.M.S. et al		7 2021	2				20			1	1			1										treatment
Searl, J. et al		26 2021	1																1					Vocal Monitor
Koyuncu, H. et al		25 2021	7.																1				diet	Total Monitor
Yasar, O.C. et al		24 2022	2					1		1			1	1	1			1			1	1		
Rajasekar, S.S		3 2022	_					1 1					1								1	1	Istm cnn	accuracy 85%
,,																							Sup vector machine	
Suppa, A. et al	2	2022	11	5		:	108												1		1		classifier	Audio recorder, LR- value
Yu, Q. et al	2	2022	8	0			40	1		1		1	1					1			1	1	SVM, accuracy 73%	27 voice features
Paulino, C.E.B. et al	2	2022	2	0			20								1	1								
Kopf, L.M. et al	1	9 2022	2	4															1		1			Compares 12 STN/12 GPI
Vojtech, J.M. et al	1	8 2022	2	0			20																1 utterance /ifi/	
Dos Santos, A.P. et al	1	7 2022	1	4	1	1	1			1	1					1							Voiss/V-RQOL	
Pah, N.D. et al	1	6 2022	5	0			50				1							1					SVM	Acuraccy 84%
Bao, G. et al	1	5 2022						1					1	1							1		PCA	SSCL, acuraccy 83%
Marchese, M.R. et al	1	4 2022	1	5						1	1	1				1			1	1				
Dao, S.V et al		3 2022																			1		GWO/LGBM	
Sapmaz, A.M. et al	1	2022	4.	3			43	1		1		1	1	1				1	1		1		Audacity	WAS
Butala, A. et al	1	1 2022	2	6				1		1	1		1	1										singing
Lim, W.S. et al	1	.0 2022	37	1														1			1		auroc	
Good, A. et al		9 2023	2	2						1	1		1	1				1						
Cabestany, J. et al		8 2023																						Editorial
Constantini, G. et al		7 2023	12	4	1		266	1	1	1	1		1	1	1			1			1	1	SVM/CFS	453 vocal features
Qiang, L. et al		6 2023	5.	5	1		55													1			1.11	Not described accoustics
Olivares, A. et al		5 2023	2	0	1					1											1	1		
Silva, J. M. S. et al		4 2023	2	0	1		20		1	1	1		1	1									_	
Abraham, E. A. et al		3 2023	1	2	1		12			1			1	1						1			6	11 accoustic parameters MI
ima, H. V. S. L. et al		2 2023	3	0			30	1	1				1	1		1								
Romero Arias, T. et al		1 2023	2		1					1		1	1	1					1	1				the Online Lab App tool?
		7:	61(23																					
		W	ithout																					

This validation is based on 7561 patients (23 papers without numbers) and 1513 controls (58 without numbers) in 98 papers from 2013 to 2023 (minus 5 reviews)

Most studies are on early and moderate cases of Parkinson's' disease. 7 papers present results of deep brain treatment

Mostly, validations in non-Al papers are: HNR F0 intensity Jitter Shimmer and VHI Also, in non-Al papers are: SNR MPT Spectrography Cepstrum analysis VRP GRBAS

Praat is used in both non-AI and some AI cases. AI is used for validation in 24 papers and is often based on many more parameters

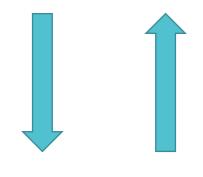
<u>Parameters</u>	<u>Total</u>
	7561 (23 without
No Patient (cases)	no.)
Prospective	25
Randomized	5
	1513 (58 without
(Case) Controls	no.)
Retrospective	6
HNR	23
SNR	8
F0 (+stnd. dv.)	40
Intensity	24
MPT	14
JITTER APS/%	29
SHIMMER APS/%	23
Spekt LTAS	9
CEPSTRUM	5
VRP	4
Telephone	3
Praat	13
VHI	25
GRBAS	10
Deep Brain.s	7
Al	24
Deep Learning	9
Laryngoscopy	6



The Search

Voice related biomarkers

Disease



Voice Change

- Up to 78% of early stage **Parkinson-patients**
- **NDD:** early diagnostic

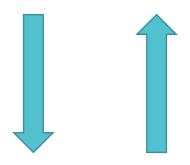






Voice related biomarkers

Disease



Voice Change

- Up to 78% of early stage **Parkinson-patients**
- **NDD:** early diagnostic
- Alzheimer, cognitive diseases: often combined with language, verbal fluency, word finding difficulties, semantic errors
- MS & Reumatoid arthritis: voice and phonatory behaviour (DBS!): >> articulation, respiration & prosody
- Mental health & emotions: acoustics & linguistics
- Cardiovasc & diabetes
- Covid, resp condition





The Search

Voice related biomarkers

Disease



Voice Change (lit)

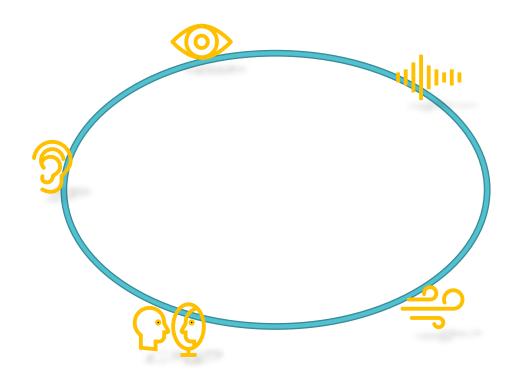


Voice / glottal function





Multidimensionality



Acoustics alone is not sufficient.
Linguistics, semantics & vocabulary, ... do not reflect glottal function.



Innovation

Clinical BM_Glottal Function

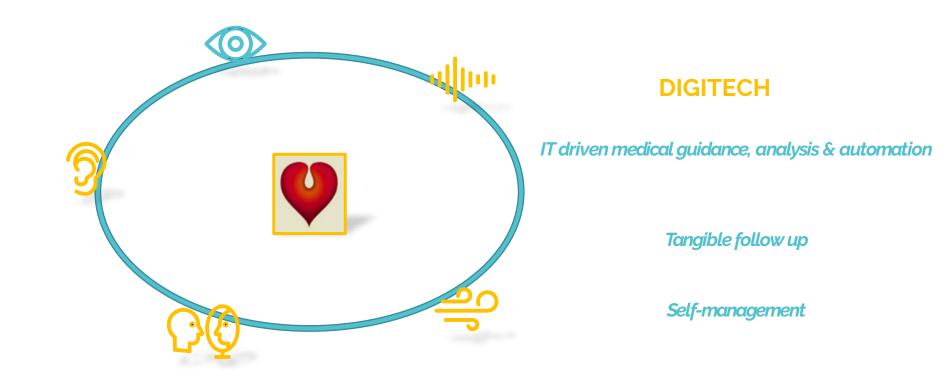
MEDTECH

Screening

Triage

Assisted diagnosis & therapy

Severity assessment



Personalized guidance

Virtual testing
Available/Affordable/Accessible





Towards a preliminary clinical protocol, suggested by the UEP BMC

1st target population

Parkinson's disease, other neuromuscular disorders

The sample

- sustained open vowel /a/ on a comfortable loudness and pitch
- sustained vowel /a/ as loud as possible (but no shouting)
- sustained vowel /a/ as silent as possible (but no whisper)
- sustained vowel /a/ on the highest possible pitch (chest register- no singing)
- sustained vowel /a/ on the lowest possible pitch (chest register- no singing)
- sentence without fricatives (for Dutch "aan die bemiddeling willen we meedoen")
- phonetically balanced text passage (for Dutch "Papa en Marloes"; for German
 "Nordwind und Sonne"; for English "The Rainbow Passage", ...)



The How

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Important and easily accessible parameters with potential of leading to a biomarker for glottal closure

We suggest to keep **at least** the following parameters, considering their importance in the glottal closure:

- Perception: G & B from the GRBAS.
 In perspective of Voicing related parameters, the IINFVo perceptual rating scale might form an alternative (Eur Arch Otorhinolaryngol. 2006 May;263(5)
- Acoustics: NHR, Shimmer, dB, Voicing quantification parameters (Fractal Dimension, Normalized Mutual Information).
- Visualization: since this requires the intervention of a clinician, we ignore this for the moment. However, visualization of glottal closure can be used as a reference for the glottal insufficiency biomarker in a clinical trial.
- Aerodynamics: MPT, glottal closure quotient, the flow declination rate via inverse filtering (Sopran, Aalto Aparat, OnlineApp Lab tool)
- Self-assessment: specific items of the VHI linked to glottal closure/voicing.





Conclusions

Voice is different from speech and language

Voice is multidimensional

Glottal closure is vitally important

Determining a BM should be simple and straightforward, affordable, largely accessible and available

Presumably AI is needed to determine the weight of the different dimensions and to obtain an accurate BM

The committee decided to work on defining (and ultimately testing) biomarkers that reveal glottal function. And as the three most important and relevant functions of the glottis are to facilitate ventilation, to provide phonation and to protect the airways, the focus primarily lies on the glottal closure. Good glottal closure is not only necessary for a good voice, but also to prevent aspiration.





Conclusions

Invitation

Reflections on the suggested protocol?

Volunteers with access to NDD/parkinson's disease to participate in a clinical trial?

New members for the UEP BMC?





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