

INTRODUCTION

Characterizing speech motor performance in dysarthria important for diagnosis and treatment

- One way to assess motor control over different levels of speech production is to estimate the stability of movement patterns.
- Kinematic measures of speech motor variability (EMA, strain-gauge transducers) indicate changes in dysarthric speakers, but are expensive and invasive.
- Acoustically based measures also promising in signalling presence and severity of dysarthria [1].

Aim of the study

Evaluate speaking conditions and acoustic parameters of variability measures for their suitability to diagnose and classify dysarthria.

METHODOLOGY

Speakers

- 23 speakers with Parkinson's disease and mild to moderate **hypokinetic dysarthria** (HD): 18 male, 5 female, age 40-81, M=66.6, SD=10.6.
- 9 speakers with various neurological diseases and mild to severe **ataxic dysarthria** (AD): 6 male, 3 female, age 37-70, M=49.0, SD=11.8.
- 27 age-matched **control speakers** (AMC): 16 male, 11 female, age 35-80, M=57.4, SD=13.9.

Procedure

- Stimuli: Repeat the phrase "Tony knew you were lying in bed" 20 times
- Six speaking conditions: **Habitual** rate, **Slow** rate, **Fast** rate, **Increased Length (IL)** "One two three Tony knew you were lying in bed five six seven", **Increased Complexity (IC)** "I heard that Tony knew you were lying in bed this Sunday morning", and **Dual** task (during spiral drawing).

DATA ANALYSIS

Experimental setup

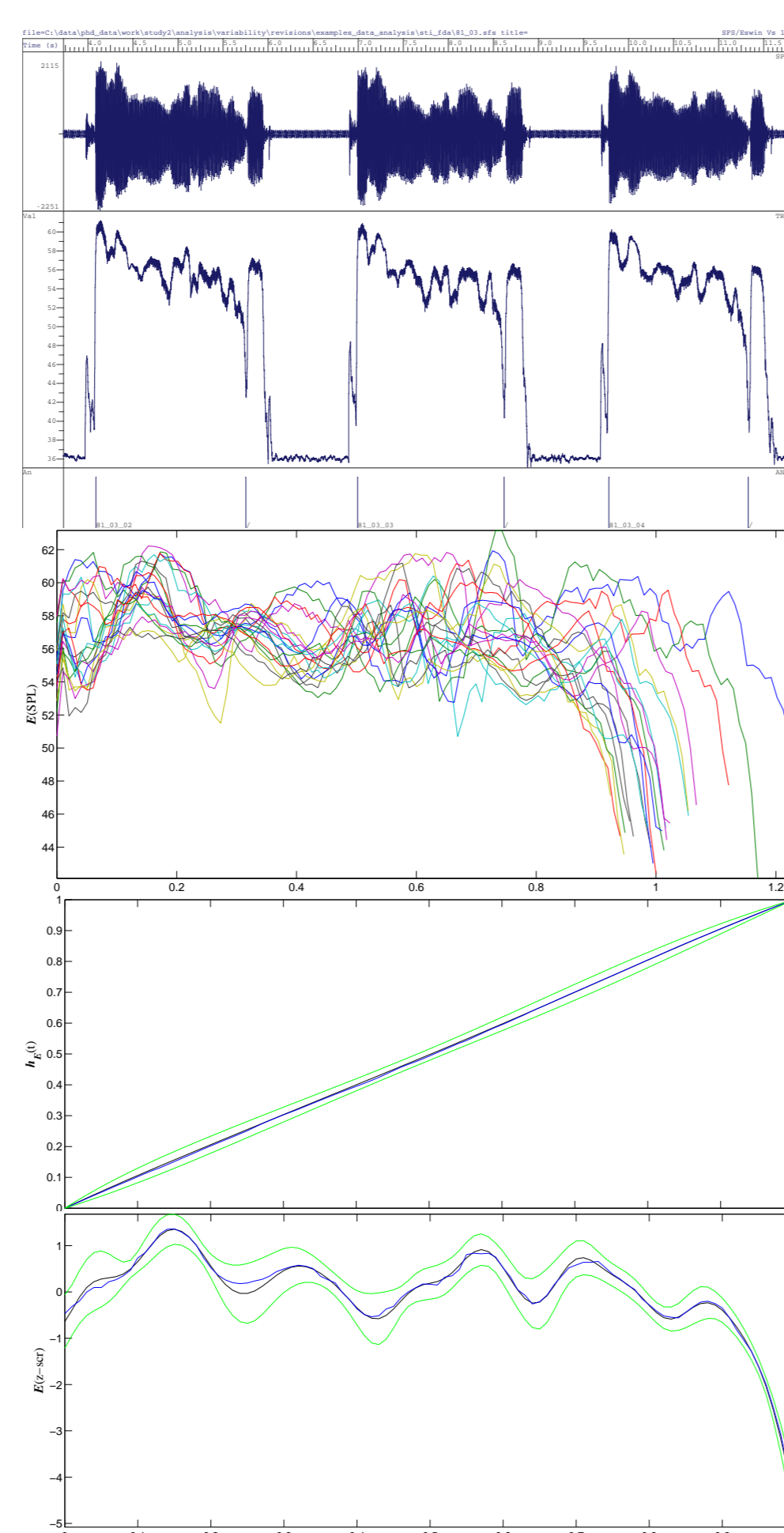
Audio data collected with a portable audio-recorder and head-mounted microphone.

Variability analysis

- **Annotation** of phrase repetitions.
- **Extraction** of contours Sound Pressure Level (SPL), Fundamental Frequency (F0), First Formant (F1), and Second Formant (F2).
- **Processing** of contours with Functional Data Analysis to obtain spatial variability (SV), temporal variability (TV), and the spatiotemporal index (STI) [2].

Statistical analyses

- 72 variables obtained [4 speech parameters X 6 speaking conditions X 3 variability measures].
- Data reduction with **Principal Component Analysis**; extraction of oblique rotated factors [3].
- **Logistic Regression** to analyse the relationship between the extracted factors and outcome (dysarthria / unaffected; dysarthria type) [4].



RESULTS

	Hypokinetic vs Controls				Ataxic vs Controls				Dysarthria vs Controls				Hypokinetic vs Ataxic			
	Principal Component Analysis															
Number of Factors	16				15				16				16			
% variance explained	86.6%				89.7%				85.9%				91.6%			
	Logistic Regression															
Block 0 Constant	B	S.E.	Sig.	Exp(B)	B	S.E.	Sig.	Exp(B)	B	S.E.	Sig.	Exp(B)	B	S.E.	Sig.	Exp(B)
Block 1 Model fit	- 2LL: 40.11 (from 68.99) Nagelkerke R ² = .586				- 2LL: 22.99 (from 40.49) Nagelkerke R ² = .570				- 2LL: 38.48 (from 81.37) Nagelkerke R ² = .690				- 2LL: 30.732 (from 38.02) Nagelkerke R ² = .293			
Block 1 Constant	B	S.E.	Sig.	Exp(B)	B	S.E.	Sig.	Exp(B)	B	S.E.	Sig.	Exp(B)	B	S.E.	Sig.	Exp(B)
	.125	.444	.778	1.134	1.631	.593	.006	5.109	1.283	.633	.043	3.606	1.096	.458	.017	2.993
Classification Table		HD	AMC	% correct		AD	AMC	% correct		DYS	AMC	% correct		HD	AD	% correct
	HD	18	5	78.3	AD	6	3	66.7	DYS	28	4	85.2	HD	21	2	91.3
	AMC	23	4	85.2	AMC	27	0	100	AMC	4	23	87.5	AD	6	3	33.3
Overall % correct	82.0				91.7				86.4				75.0			
Contributing Factors / Variables	1 / 5				2 / 16				6 / 31				1 / 5			
Prominent Variables	{STI,SV,TV}_SPL_IC TV_F1_{Slow,IC}				{STI,SV,TV}_SPL_Slow SV_SPL_{Hab,IL,IC} {STI,SV}_F0_{Hab,Slow,IC} TV_F1_{Slow,IC}				Trends: {STI,SV}_SPL {STI,SV,TV}_F0 F1_{Hab,Slow} F2_{IL,Dual}				SV_SPL_{Hab, Slow,Fast,IL,Dual}			

DISCUSSION

Principal Component Analysis

- Grouping 72 variables into relatively high number of factors (15-16).
- First 2 factors explain only 36 - 41% of total variance.

Logistic Regression

- Using PCA rotated factors as predictors resulted in improved logistic models.
- Each model contained at least 1 significant factor that improved the models.

Classification

- Classifications HD vs AMC and DYS vs AMC reasonably successful.
- AD vs AMC: 1 in 3 are classified as false negatives.
- HD vs AD: many AD speakers classified as HD.
- Possibly due to low sample size and varying speaker profiles in the AD group.

Parameter Selection

- HD vs AMC: SPL variability higher in HD group during repetition of phrase in IC speaking condition.
- AD vs AMC: increased SPL and F0 variability in Hab, Slow, and IC conditions.
- DYS vs AMC: difficult to select small number of diagnostic parameters; increased variability across all acoustic parameters and most speaking conditions.
- HD vs AD: increased spatial variability of SPL in AD group.

Conclusions

- Acoustic measures of variability may be used to **signal dysarthria**: HD (SPL, F1) and AD (SPL, F0, F1).
- ...and to **distinguish dysarthria types** (SV of SPL).
- Most robust overall: **Spatial Variability of Sound Pressure Level in Slow and Increased Complexity** conditions.
- Demonstrates added value of Functional Data analysis to STI.

Limitations

- Low sample sizes (AD group) and missing data (F2 contours).
- Different underlying etiologies in speakers with ataxic dysarthria.
- HD and AD group not comparable in severity (based on intelligibility).

REFERENCES

- [1] van Brenk, F., & Lowit, A. (2012). The relationship between acoustic indices of speech motor control variability and other measures of speech performance in dysarthria. *Journal of Medical Speech Language Pathology*, 20(4), 24-29.
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- [4] Reed, P., & Wu, Y. (2013). Logistic regression for risk factor modelling in stuttering research. *Journal of fluency disorders*, 38(2), 88-101.

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