Natural speech as a digital biomarker in preclinical Alzheimer's disease: Usability and reliability of a remote burst speech assessment

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Background

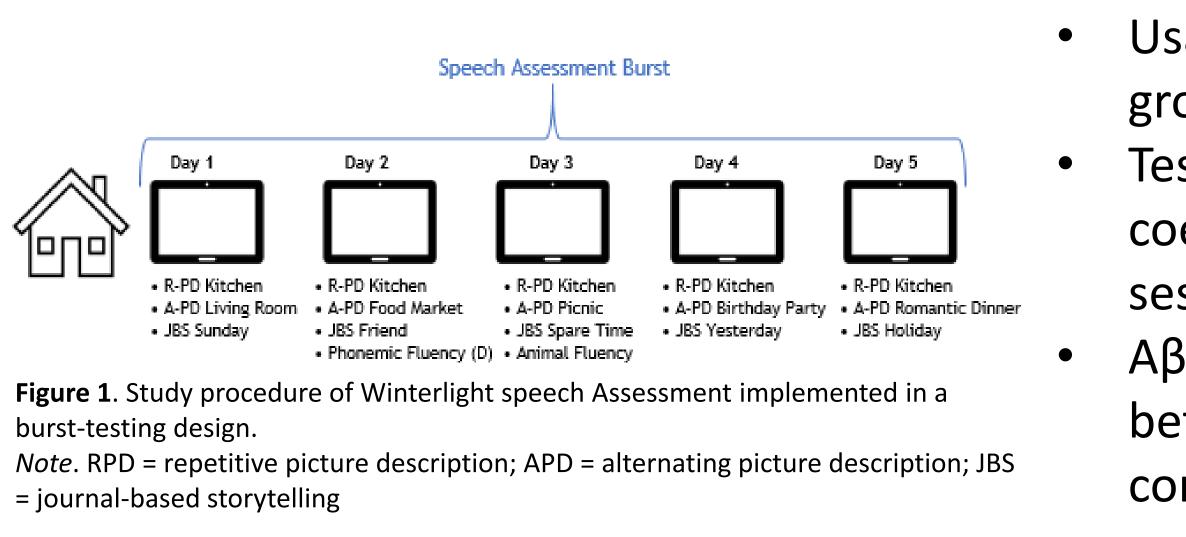
- Language difficulties are often reported as one of the earliest symptoms in Alzheimer's disease $(AD)^{1-3}$, but evidence is inconclusive about the sensitivity of traditional neuropsychological language tests for early detection⁴⁻⁶.
- Fine-grained speech analysis offers the potential for capturing subtle cognitive deficits in early-stage AD.^{7,8}
- As such, digital recordings of natural speech are a promising **digital biomarker**, particularly when measured in **bursts** (repeated short measurements).^{9,10}

Aim

Evaluate the **usability** of a tablet-based speech assessment and examine the **test-retest** reliability of acoustic speech features and its association with amyloid-beta (AB) **pathology** (+/-) in cognitively normal (CN) adults.

Methods

- Participants: 50 cognitively normal (CN) Dutch-speaking adults from the Alzheimer Center Amsterdam (Table 1).
- Test-retest design within 2-3 week interval, in an at-home setting.
- Measures: 1) Winterlight Speech Assessment on a tablet, implemented in a burst-design (Figure 1), 2) System usability scale (SUS): 5point Likert scale to evaluate usability of the speech assessment.
- Acoustic speech features were extracted from Note. SD = standard deviation, * indicates p-value < 0.05 speech recordings (e.g. silent pauses, fundamental frequency, jitter, shimmer).



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Table 1. Participant characteristics

	Amyloid-beta	Amyloid-beta
	positive (N=23)	negative (N=27)
Demographics		
Age, years, mean ± SD	69.61 ± 6.34	67.30 ± 6.03
Female gender, n (%)	13 (56.5)	16 (59.3)
Education, years, mean ± SD	15.22 ± 4.61	15.30 ± 2.97
Own tablet (iOS), n (%)	6 (26.1)	15 (55.5)
Amyloid-beta biomarkers		
Cerebrospinal Fluid, n (%)	3 (13.0)	2 (7.4)
Positron Emission Tomography, n (%)	20 (87.0)	25 (92.6)
Cognitive measures		
MMSE, mean ± SD*	28.83 ± 1.07	29.52 ± 0.70
CDR, mean ± SD	0±0	0±0
	1 . 0.05	

Statistical analyses

Usability: Differences in SUS-scores between groups: Wilcoxon rank sum test.

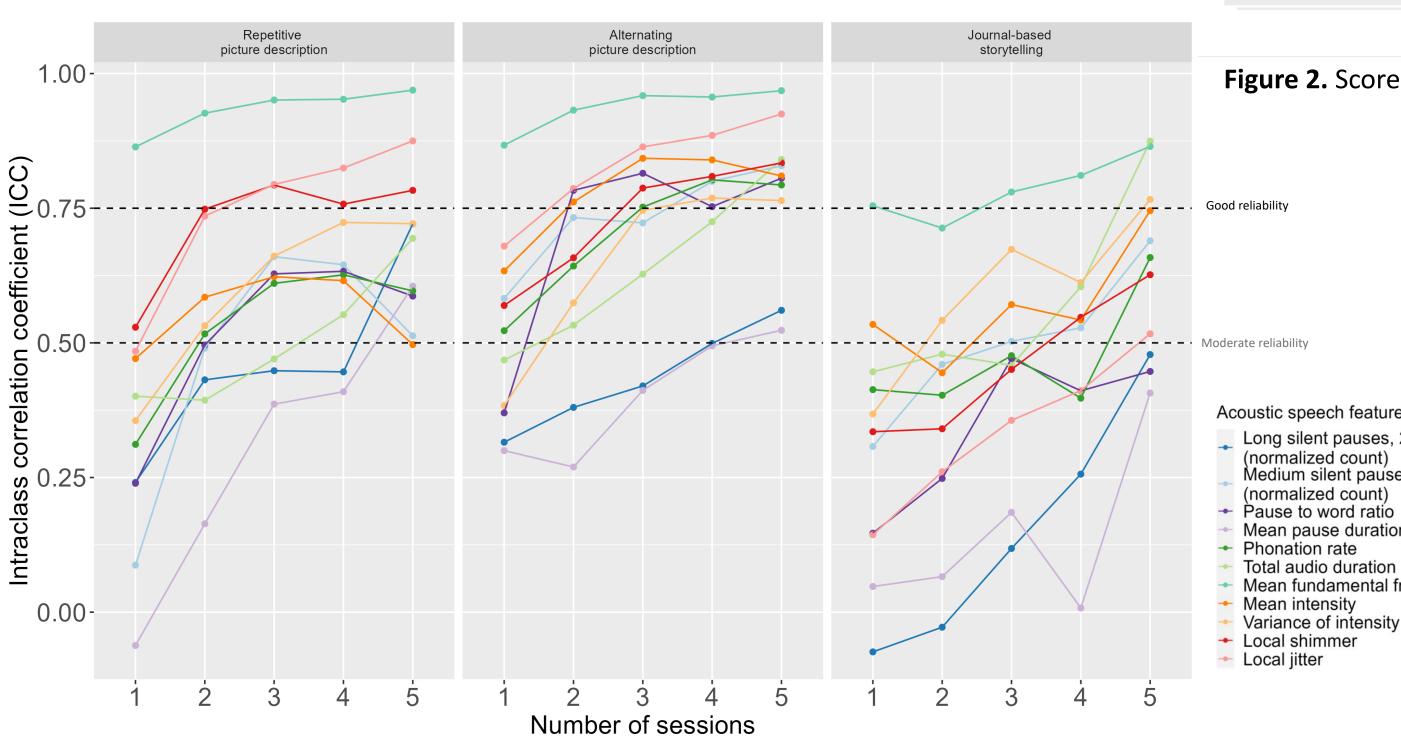
Test-retest reliability: Intraclass correlation coefficients (ICC) in cumulative numbers of sessions per subtask, for each feature. Aβ-pathology: Differences in acoustic features between $A\beta$ + and $A\beta$ - groups: linear models (LMs) corrected for age, sex, education and MMSE.

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Usability: The average SUS score was 86 (SD = 9.88). The A β + group evaluated usability as good (M=83±10), while the Aβ- group rated usability to be excellent (M= 88 ± 9 , p = 0.019; Figure 2).



A β -pathology: The A β +-group showed more medium pauses in 3- and 4-session-bursts of journal-based storytelling than the Aβ-group $(_{3-\text{session-bursts:}} \beta = -0.07, C = -0.12 - -0.01, p = 0.016;$ 4-session-bursts: β=-0.06, C/=-0.11–-0.01, p=0.029; Figure 4.). For none of the other number of sessions, subtasks, or acoustic features group differences were found (p's > .05).



¹Valech et al., 2018; ²Montembeault et al., 2022, ³Slegers et al., 2018, ⁴Papp et al., 2016; ⁵Vonk et al., 2020; ⁶Mueller et al., 2021, ⁷Petti et al. (2020); ⁸Martínez-Nicolás (2021), ⁹Moore et al (2022), ¹⁰Nicosia et al. (2022)

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Figure 3. Intraclass correlation coefficients for test-retest reliabilities (2-3 week interval) for acoustic speech features in cumulative numbers of burst of repetitive-PD, alternating-PD and journal-based storytelling.

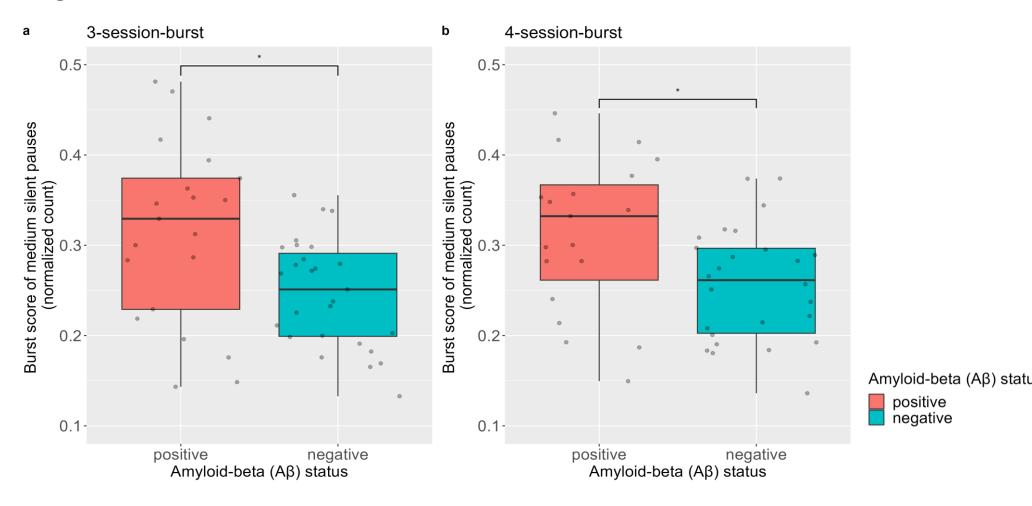


Figure 4. Medium silent pauses (normalized count) in $A\beta$ + and $A\beta$ - individuals in a) 3-session-bursts and b) 4-session-bursts of journaling question storytelling.

Conclusion

- Usability: Remote burst assessment of speech is feasible in CN older adults.
- Test-retest reliability: Burst assessments enhance test-retest reliability compared to one-session measures
- Aβ-pathology: Burst assessments of acoustic speech features are promising to find differences in speech acoustics between $A\beta$ + and $A\beta$ -negative individuals.
- These results suggest remote burst assessment holds promise for detecting subtle acoustic speech changes in the earliest AD stages.



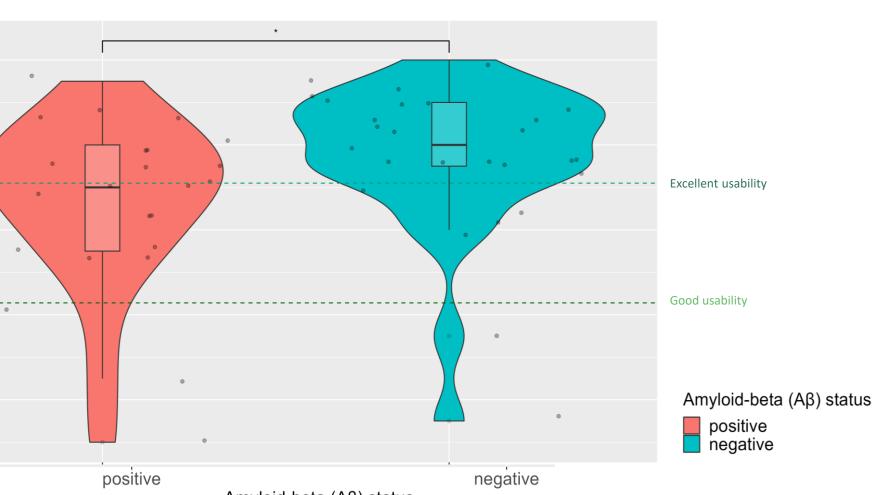


Figure 2. Scores on the system usability scale (SUS) in $A\beta$ + and $A\beta$ - individuals.

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Test-retest reliability: ICCs increased in more-sessionbursts compared to onesession measures for all acoustic features in all speech subtasks (Figure 3).

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