REPORT

MemoTag: Voice-Based Cognitive Decline Pattern Detection

Proof of Concept Report

6 Objective

To build a basic proof-of-concept pipeline that processes anonymized voice samples and extracts early indicators of cognitive stress or decline using audio features and unsupervised machine learning.

Methodology

1. Audio Preprocessing + Transcription

- Voice clips were extracted from a ZIP archive and preprocessed.
- · Transcription was performed using OpenAI's Whisper model for robust handling of realistic, noisy input.

2. Feature Extraction

From each audio file, the following cognitive-linguistic markers were extracted:

- Speech Rate (words per second)
- Number of Hesitations (uh, um, etc.)
- Number of Pauses (detected using silent intervals via librosa)
- Pitch Mean & Pitch Variability (measured from pitch tracking)

3. Unsupervised Anomaly Detection

- Features were standardized using StandardScaler.
- Isolation Forest was applied to detect anomalies potentially indicative of cognitive impairment.
- Outliers were flagged based on abnormal speech patterns (e.g., slower speech, more pauses, flatter pitch).

Key Observations

- Anomalous voice samples often showed:
 - Lower speech rate
 - Higher pause/hesitation frequency
 - Reduced pitch variability

REPORT

• These align with early linguistic indicators of cognitive decline such as reduced fluency and impaired lexical access.

Technologies Used

- Python, Google Colab
- whisper , librosa , sklearn , matplotlib , seaborn , pandas

Sample Output

Filename	Speech Rate	Pauses	Hesitations	Pitch Mean	Pitch Std
sample3.wav	2.82	0	0	1312.11	987.10
(Flagged as an outlier by Isolation Forest)					

Next Steps

- Introduce **structured tasks** (e.g., sentence completion, memory recall prompts).
- Extract NLP-based features like word substitutions or repetition.
- Combine audio features with **demographic or clinical labels** for supervised learning.
- Validate with expert-annotated cognitive assessments.

REPORT 2