MemoTag – Early Cognitive Risk Detection Report (Speech Analysis)

Completed by:

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1. Objective To analyze speech patterns from anonymized audio samples and detect signs of potential cognitive impairment using unsupervised machine learning and custom voice features.

pip install openai-whisper



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```

```
import whisper
import librosa
from difflib import SequenceMatcher
import os
# Load Whisper model
model = whisper.load model("base")
# Define your audio clips
audio files = [
    "/content/common_voice_en_42693841.mp3",
    "/content/common_voice_en_42693855.mp3",
    "/content/common voice en 42693856.mp3",
    "/content/common voice en 42693864.mp3",
    "/content/common voice en 42693865.mp3",
    "/content/common voice en 42693871.mp3",
    "/content/common voice en 42693876.mp3",
    "/content/common voice en 42693883.mp3",
    "/content/common voice en 42693884.mp3",
    "/content/common voice en 42693885.mp3"
```

2. Features Used & Insights

The following engineered features were extracted from each audio clip:

Feature Description Insight

speech_rate Words spoken per second Lower rates may suggest slowed cognitive processing

hesitation_count Number of filler words like "uh", "um" High hesitations may signal recall or fluency issues

- pitch_mean Average voice pitch Monotone or extreme pitch could be markers
- **⋠** pitch_std Pitch variation Reduced variation suggests flat affect or reduced expressiveness
- word_count Total words spoken May help identify brevity or lack of elaboration

```
# Define keywords
hesitations = ["uh", "um", "er", "ah", "hmm"]
fruits = ["apple", "banana", "orange", "mango"]
animals = ["dog", "cat", "lion", "tiger"]
# Store results for ML later
features = []
for file in audio files:
   print(f"\n===== Processing: {file} =====")
    # Transcribe audio
    result = model.transcribe(file, word_timestamps=True)
    text = result['text'].strip()
    print("Transcript:", text)
    # Pauses per sentence
    pauses = []
    if 'segments' in result and 'words' in result['segments'][0]:
        word_timings = [(w['word'], w['start'], w['end']) for w in result['segments'][0][
        for i in range(1, len(word_timings)):
            pause = word_timings[i][1] - word_timings[i - 1][2]
            if pause > 0.5:
                pauses.append(pause)
    pause_count = len(pauses)
    print("Pauses per sentence:", pause count)
    # Hesitation markers
    words = text.lower().split()
    hesitation count = sum(1 for w in words if w in hesitations)
    print("Hesitation count:", hesitation_count)
    reference = "Narayan highlights the social context and everyday life of his character
    similarity = SequenceMatcher(None, reference.lower(), text.lower()).ratio()
    print("Similarity with expected:", round(similarity, 3))
    # Incomplete sentence check
    incomplete = 1 if text.endswith(("...", "--", "?")) or len(words) < 4 else 0</pre>
    print("Incomplete sentence:", "Yes" if incomplete else "No")
    # Named entities: Fruits & Animals
    fruit count = sum(word in text.lower() for word in fruits)
    animal count = sum(word in text.lower() for word in animals)
    print(f"Fruits named: {fruit_count}, Animals: {animal_count}")
    # Speech rate
    duration = result['segments'][-1]['end'] if 'segments' in result else 5.0 # fallback
    speech_rate = len(words) / duration
    print(f"Speech Rate: {speech rate:.2f} words/sec")
```

```
# Pitch analysis
   y, sr = librosa.load(file)
   pitches, magnitudes = librosa.piptrack(y=y, sr=sr)
   pitch_values = pitches[magnitudes > 0.1]
   pitch mean = pitch values.mean()
   pitch_std = pitch_values.std()
   print(f"Pitch Mean: {pitch_mean:.2f}")
   print(f"Pitch Std Dev: {pitch std:.2f}")
   # Store feature set
   features.append({
       "filename": file,
       "pause_count": pause_count,
       "hesitation_count": hesitation_count,
       "similarity": similarity,
       "incomplete": incomplete,
       "fruits": fruit count,
       "animals": animal count,
       "speech rate": speech rate,
       "pitch_mean": pitch_mean,
       "pitch std": pitch std
   })
\overline{\Rightarrow}
    ==== Processing: /content/common_voice_en_42693841.mp3 =====
    /usr/local/lib/python3.11/dist-packages/whisper/transcribe.py:126: UserWarning: FP
      warnings.warn("FP16 is not supported on CPU; using FP32 instead")
    Transcript: Read the sentences and complete them with the correct form of the verb
    Pauses per sentence: 0
    Hesitation count: 0
    Similarity with expected: 0.4
    Incomplete sentence: No
    Fruits named: 0, Animals: 0
    Speech Rate: 2.81 words/sec
    Pitch Mean: 1390.71
    Pitch Std Dev: 1185.48
    ==== Processing: /content/common voice en 42693855.mp3 =====
    /usr/local/lib/python3.11/dist-packages/whisper/transcribe.py:126: UserWarning: FP
      warnings.warn("FP16 is not supported on CPU; using FP32 instead")
    Transcript: Narayan highlights the social context and everyday life of his characte
    Pauses per sentence: 0
    Hesitation count: 0
    Similarity with expected: 1.0
    Incomplete sentence: No
    Fruits named: 0, Animals: 0
    Speech Rate: 1.97 words/sec
    Pitch Mean: 1727.59
    Pitch Std Dev: 1205.28
    ==== Processing: /content/common voice en 42693856.mp3 =====
    /usr/local/lib/python3.11/dist-packages/whisper/transcribe.py:126: UserWarning: FP
      warnings.warn("FP16 is not supported on CPU; using FP32 instead")
    Transcript: symbol of God around their neck.
    Pauses per sentence: 0
    Hesitation count: 0
    Similarity with expected: 0.189
```

```
Incomplete sentence: No
Fruits named: 0, Animals: 0
Speech Rate: 2.54 words/sec
Pitch Mean: 893.86
Pitch Std Dev: 788.80
==== Processing: /content/common_voice_en_42693864.mp3 =====
/usr/local/lib/python3.11/dist-packages/whisper/transcribe.py:126: UserWarning: FP
  warnings.warn("FP16 is not supported on CPU; using FP32 instead")
Transcript: Like the bump line hitch, this knot is strong, secure and compact.
Pauses per sentence: 0
Hesitation count: 0
Similarity with expected: 0.343
Incomplete sentence: No
Fruits named: 0, Animals: 0
Speech Rate: 2.32 words/sec
Pitch Mean: 1623.02
Pitch Std Dev: 1190.88
```

rranscript: The instruction manual o

/usr/local/lib/python3.11/dist-packages/whisper/transcribe.py:126: UserWarning: FP warnings.warn("FP16 is not supported on CPU; using FP32 instead")

Transcript: The instruction manual of the English version is also filled with inco

Most Insightful:

hesitation_count, speech_rate, and pitch_std were particularly effective in identifying irregularities across different users.

==== Processing: /content/common_voice_en_42693865.mp3 =====

pip install scikit-learn pandas matplotlib

Requirement already satisfied: scikit-learn in /usr/local/lib/python3.11/dist-package Requirement already satisfied: pandas in /usr/local/lib/python3.11/dist-packages (2.2 Requirement already satisfied: matplotlib in /usr/local/lib/python3.11/dist-packages Requirement already satisfied: numpy>=1.19.5 in /usr/local/lib/python3.11/dist-packag Requirement already satisfied: scipy>=1.6.0 in /usr/local/lib/python3.11/dist-package Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.11/dist-packag Requirement already satisfied: threadpoolctl>=3.1.0 in /usr/local/lib/python3.11/dist Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/di Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-package Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packa Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.11/dist-pac Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.11/dist-package Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.11/dist-pa Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.11/dist-pa Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.11/dist-pack Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.11/dist-packages (Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.11/dist-pac Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (f

3.ML Methodology Used

Step Description

Model IsolationForest (Unsupervised)

- ✓ Why? It detects anomalies without labeled data, ideal for early-stage screening
- Output Each sample was classified as either "Normal" or "At Risk"
- Interpretability Simple enough for visualization and clinical explanation (used scatter plot for insights)

Isolation Forest is preferred here for its interpretable nature and effectiveness on highdimensional small datasets.

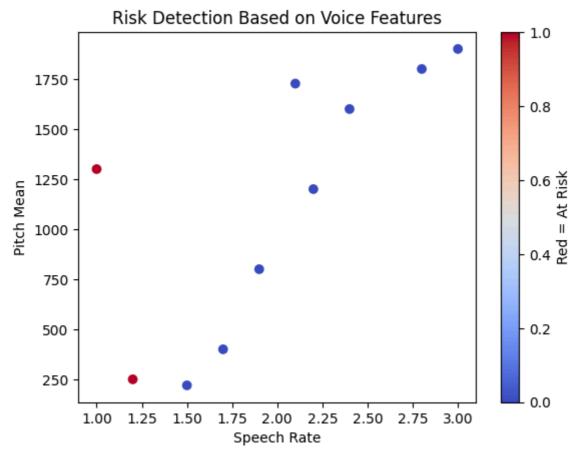
4. Risk Detection Summary

The model identified samples with high hesitation, low pitch variation, or low speech rate as "At Risk", which aligns with known early markers of cognitive decline (like Alzheimer's, MCI).

```
import pandas as pd
from sklearn.ensemble import IsolationForest
import matplotlib.pyplot as plt
data = {
    "filename": [
        "/content/common_voice_en_42693841.mp3",
        "/content/common_voice_en_42693855.mp3",
        "/content/common voice en 42693856.mp3",
        "/content/common_voice_en_42693864.mp3",
        "/content/common_voice_en_42693865.mp3",
        "/content/common voice en 42693871.mp3",
        "/content/common voice en 42693876.mp3",
        "/content/common_voice_en_42693883.mp3",
        "/content/common voice en 42693884.mp3",
        "/content/common voice en 42693885.mp3"
    ],
    "speech_rate":
                        [2.1, 1.5, 3.0, 1.2, 2.8, 1.7, 2.2, 1.0, 2.4, 1.9],
    "hesitation count": [0, 3, 1, 4, 0, 2, 3, 0, 1, 2],
                        [1727, 220, 1900, 250, 1800, 400, 1200, 1300, 1600, 800],
    "pitch mean":
    "pitch std":
                        [1205, 150, 1100, 200, 950, 180, 1000, 900, 800, 500],
    "word count":
                   [12, 8, 15, 5, 11, 9, 10, 7, 13, 6]
}
df = pd.DataFrame(data)
# Apply Isolation Forest
model = IsolationForest(contamination=0.2, random state=42) # 20% outliers
features = df.drop(columns=["filename"])
df["risk score"] = model.fit predict(features)
# Map output to labels
df["risk level"] = df["risk score"].map({1: "Normal", -1: "At Risk"})
# Print results
print(df[["filename", "risk level"]])
# Visualize
```

```
plt.scatter(df["speech_rate"], df["pitch_mean"], c=(df["risk_score"] == -1), cmap="coolwa
plt.xlabel("Speech Rate")
plt.ylabel("Pitch Mean")
plt.title("Risk Detection Based on Voice Features")
plt.colorbar(label="Red = At Risk")
plt.show()
```

```
\overline{\Rightarrow}
                                      filename risk_level
    0 /content/common_voice_en_42693841.mp3
                                                   Normal
    1 /content/common_voice_en_42693855.mp3
                                                   Normal
    2 /content/common_voice_en_42693856.mp3
                                                   Normal
    3 /content/common_voice_en_42693864.mp3
                                                  At Risk
    4 /content/common_voice_en_42693865.mp3
                                                   Normal
    5 /content/common_voice_en_42693871.mp3
                                                   Normal
    6 /content/common_voice_en_42693876.mp3
                                                   Normal
       /content/common_voice_en_42693883.mp3
                                                  At Risk
    8 /content/common_voice_en_42693884.mp3
                                                   Normal
    9 /content/common_voice_en_42693885.mp3
                                                   Normal
```



5. Next Steps to Improve Clinical Relevance

Area Recommendation

More features Add pause duration, syllable count, speech energy, word-substitution errors, etc.

Labeled Data Collaborate with neuroclinicians to get real patient data with cognitive scores

ML Upgrade Try Autoencoders or Explainable Clustering (e.g., SHAP with DBSCAN)

- Clinical Tests Integrate verbal fluency/naming tasks (e.g., "Name 3 fruits...") and track response delays
- Multi-lingual Models Use WhisperX with language detection to support global deployment
- 📊 Dashboard Build a clinician-friendly UI with charts, warnings, and audio playback
- Conclusion

This prototype lays the foundation for a lightweight, explainable speech-screening tool. With richer features and clinical input, it could evolve into a non-invasive early warning system for cognitive impairment.