

DEVELOPMENT OF SIGN LANGUAGE DETECTION AND CONVERSION SYSTEM USING MEDIAPIPE

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Objective

- ① Use MediaPipe for accurate, real-time gesture recognition and conversion to speech.
- ② Translate spoken language (based on ASL) into visual sign language for deaf individuals.
- ③ Optimize for accurate, responsive communication in real time.
- ④ Ensure compatibility across devices for easy adoption in various settings.

Methodology Flowchart

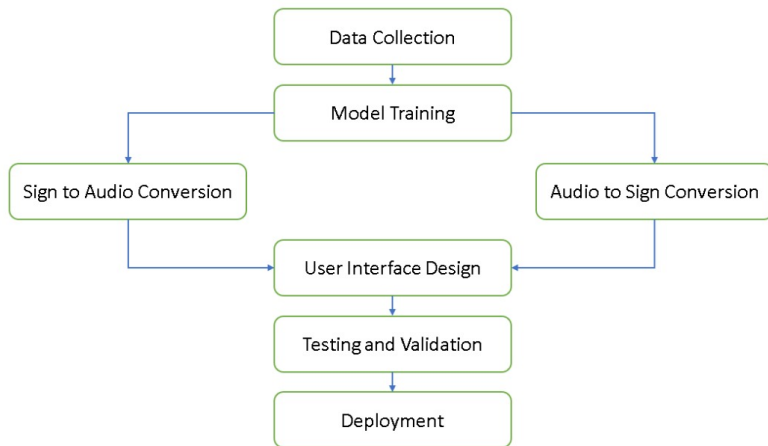


Figure: Block Diagram of Sign Language Detection-Conversion System

Software and Hardware Requirements

Category	Requirement	Use
Software	Python	Developing the machine learning models and system logic.
Python-Library file	<u>MediaPipe</u>	Real-time hand tracking and gesture recognition framework.
Python-Library file	OpenCV	Image and video processing for capturing and <u>analyzing</u> sign language gestures.
Python-Library file	<u>PyAudio</u>	For audio processing and conversion (used in audio-to-sign language conversion).
Python-Library file	<u>Tkinter</u>	For building a user-friendly interface for the system.
Hardware	Laptop with GPU	Regarding executing the machine learning model

Figure: Software and Hardware Requirements

Preliminary Result

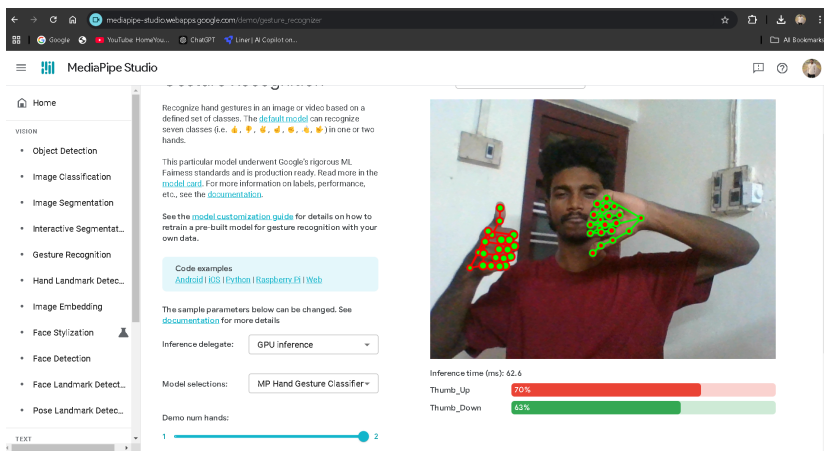


Figure: Web Demo of Basic Gesture Recognition in MediaPipe

Preliminary Result

```
app.py
4 import copy
5 import argparse
6 import itertools
7 from collections import Counter
8 from collections import deque
9
10 import cv2 as cv
11 import numpy as np
12 import mediapipe as mp
13
14 from utils import cvFpsCalc
15 from model import KeyPointClassifier
16 from model import PointHistoryClassifier
17
18
19 def get_args():
20     parser = argparse.ArgumentParser()
21
22     parser.add_argument('--device', type=int, default=0)
23     parser.add_argument('--width', help='cap width')
24     parser.add_argument('--height', help='cap height')
25
26     parser.add_argument('--use_static_image_mode',
27     parser.add_argument('--min_detection_confidence',
28         help='min_detection_confidence',
29         type=float,
```

INFO: Created TensorFlow Lite XNNPACK delegate for CPU.
INFO: Created TensorFlow Lite XNNPACK delegate for CPU.
WARNING: All log messages before absl:InitializeLog() is called are written to STDERR
WARNING:00:00:1730491730.121956 3716 inference_feedback_manager.cc:114] Feedback manager requires a model with a single signature inference. Disabling support for feedback tensors.
WARNING:00:00:1730491730.167661 3716 inference_feedback_manager.cc:114] Feedback manager requires a model with a single signature inference. Disabling support for feedback tensors.
C:\Users\shafe\AppData\Local\Programs\Python\Python312\Lib\site-packages\google\protobuf\symbol_database.py:55: UserWarning: SymbolDatabase.GetPrototype() is deprecated. Please use message_factory.GetMessageClass() instead. SymbolDatabase.GetPrototype() will be removed soon.
warnings.warn('SymbolDatabase.GetPrototype() is deprecated. Please '

FPS:19.06 Right:Open
Finger Gesture:Stop

Figure: Basic Gesture Recognition Executed in MS VS CODE

PRELIMINARY RESULTS

- The storage structure of the project directory for collecting, training and experimenting the hand gesture models.

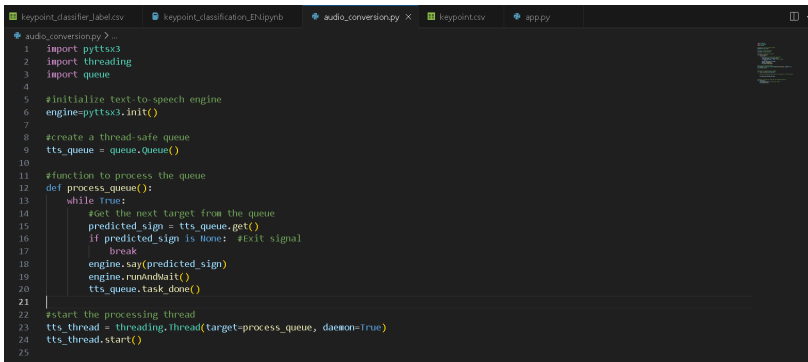
```
| app.py
| keypoint_classification.ipynb
| point_history_classification.ipynb
|
|--model
| | |--keypoint_classifier
| | | | |--keypoint.csv
| | | | |--keypoint_classifier.hdf5
| | | | |--keypoint_classifier.py
| | | | |--keypoint_classifier.tflite
| | | | |--keypoint_classifier_label.csv
| | |
| | |--point_history_classifier
| | | |--point_history.csv
| | | |--point_history_classifier.hdf5
| | | |--point_history_classifier.py
| | | |--point_history_classifier.tflite
| | | |--point_history_classifier_label.csv
| |
|
|--utils
| |--cvfpscalc.py
```

Figure: Directory Contents for Data Collection and Processing

INTERIM RESULTS:

1. Hand Sign-to-Audio Conversion (audio_conversion.py)

- Library used: pyttsx3 is used to convert the text to speech output,
- detected hand sign as text are announced as audio in real time

A screenshot of a Jupyter Notebook interface with a dark theme. The top bar shows several tabs: 'keypoint_classifier_label.csv', 'keypoint_classification_EN1.py', 'audio_conversion.py' (which is the active tab), 'keypoint.csv', and 'app.py'. The 'audio_conversion.py' tab contains a Python script. The script imports 'pyttsx3', 'threading', and 'queue'. It initializes a text-to-speech engine using 'pyttsx3.init()'. It creates a thread-safe queue named 'tts_queue'. It defines a function 'process_queue()' that runs in a loop, getting items from the queue and using the engine to say them. It starts a thread 'tts_thread' to process the queue in the background. The script is as follows:

```
1 import pyttsx3
2 import threading
3 import queue
4
5 #initialize text-to-speech engine
6 engine=pyttsx3.init()
7
8 #create a thread-safe queue
9 tts_queue = queue.Queue()
10
11 #function to process the queue
12 def process_queue():
13     while True:
14         #Get the next target from the queue
15         predicted_sign = tts_queue.get()
16         if predicted_sign is None: #Exit signal
17             break
18         engine.say(predicted_sign)
19         engine.runAndWait()
20         tts_queue.task_done()
21
22 #start the processing thread
23 tts_thread = threading.Thread(target=process_queue, daemon=True)
24 tts_thread.start()
25
26
```

Figure: script for audio conversion

INTERIM RESULTS:

2. Voice-to-Hand Sign Conversion (audio to handsign.py)

- Library used: speech recognition (Google Web Speech API).
- Matches the recognized text with "hand sign mapping" dictionary. Displays the corresponding hand sign image in a Tkinter window.

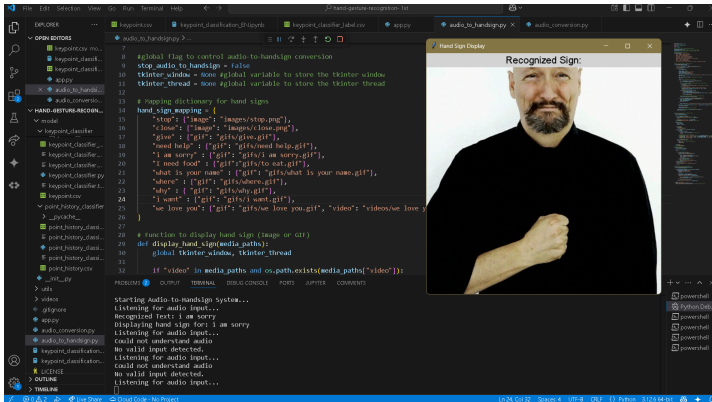


Figure: Running the Voice-to-Handsign Conversion script

INTERIM RESULTS:

- **Current Features:**

- (i) Hand Sign Recognition based on ASL:

- Utilizes Mediapipe for real-time hand landmark detection. Recognizes 15 specific hand signs:
Open (ID = 0), Close (ID=1), where is (ID=2), Ok (ID=3), I (ID=4), Want(ID=5), You(ID=6), To eat(Food)(ID=7), give(ID=8), What(ID=9), name(ID=10) I'm Sorry(ID=11) why(ID=12) How(ID=13) I need help(ID=14)

- (ii) Audio Conversions(announcing and Displaying hand gesture):

- Detected hand signs are converted into corresponding audio output using pyttsx3.
 - The captured voice is recognised using Google Web Speech API to display resp. hand sign.

INTERIM RESULTS:

(iii) Real-Time Performance:

- Processes webcam input and overlays bounding boxes, hand landmarks, and the detected hand sign label on the video feed.

(iv) Modular Structure:

- The project code is organized into separate modules:
The "audio conversion.py" handles text-to-speech conversion.
"app.py" serves as the main script for hand sign detection and system integration.

Challenges: Errors and Solutions during Interim Phase

1. FPS Drop and Unnecessary Delay :
2. RuntimeError in "audio conversion.py":queue-based system to handle TTS requests
3. Repeated Audio Output for the Same Gesture: solved by introducing TTS delay(TTS queue cleared).
4. Offline incompatibility of Google Web Speech API, used in Speech Recognition Library(VOSK, whisper by OpenAI)

Challenges: Developments after Interim Phase

1. Confidence threshold based recognition feature :

- Cause: previously Opencv was detecting the hand sign which are most close to its own understanding, increasing false positives.
- Solution: Set a default value for min. confidence threshold.
If it doesn't meet min.threshold, displays "UNKNOWN" but doesn't announce.

Challenges: Developments After Interim Phase

2. Introduced "Gesture Speed Adaption and Tracking" Algorithm:

- The system detect signing speed and dynamically adjust the confidence threshold. Faster gestures require a higher confidence threshold, reducing false positives.

How It Works:

- Track the position of key landmarks (e.g., wrist, fingers) over time.
- Calculate speed based on distance moved per frame.(Euclidean Distance formula)
- Adjust confidence thresholds or frame capture rate dynamically based on speed.(slow,normal,fast)

Challenges: Developments After Interim Phase

3. Introduced Sentence Formation using Buffer:

- Added sign sentence buffer at the beginning of main().
- Modified gesture recognition to store words in the buffer.
- Added logic to announce full sentences after a delay of 3 seconds.
- Added display of sentence buffer on the OpenCV window.

4. Added toggling ON/OFF for TTS announcement.

5. Extended to video and GIF mapping for the "audio to handsign.py" from only image mapping.

6. Improved UI tkinter window, more user friendly, with control buttons

Challenges: Developments After Interim phase

7. Customly Trained the system with American Sign Language(ASL).

- Used a Two-Key Combination for logging keypoints of Class IDs upto 99.
- allows you to log up to 100 unique gestures (class IDs 0–99) by pressing two keys in sequence (e.g., 1 + 0 for class ID 10).
- Since we have trained 15 ASL hand gestures, The total number of sentences that can be formed by the '15 ASL gestures' that I customly trained is around '3000' random sentences!

Examples: "I want you," "I need help," "I'm sorry." "What is your name?", "Why?", "How?", "You need help.", "I want you to eat.", "Why need help?".. and so on.

Controls and Key Bindings

- k = to log the keypoints
- h = to log the point history
- n = Toggle audio to handsign conversion mode
- Pressing 0 to 9 and upto 99(Class IDs) and = to logging the keypoints to class IDs certain hand landmarks
- c = to clear the sentence buffer
- s = to terminate creating sentences
- a = to toggle TTS announcement

Conclusion

- Overcame Technical Challenges :
Fixed FPS drop, TTS errors, and offline incompatibility issues. Introduced confidence thresholds and Gesture Speed Tracking for better accuracy.
- Enhanced User Experience :
Added sentence formation, TTS toggling, and video/GIF support. Improved UI with user-friendly controls.
- Custom Hand gestures trained based on ASL :
Trained 15 ASL gestures for now, enabling 3000+ sentence combinations. Examples: "I want you," "I need help," "What is your name?"

Work Distribution

SHAFEEHE	System Design and Methodology, Setting algorithm for gesture detection and audio conversions.
FIDA	Data Collection, Prepare Datasets and Model Training for Gesture Recognition.
NIDA	Integrate MediaPipe and TTS components, Design User Interface, Conducting system tests.
JASIR	Validating model accuracy, Project Coordination, User feedback.
ALL MEMBERS	Literature Review, Documentation and Presentation.

References

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THANK YOU