

Integration into 3D Printing for Image Processing using AI ML

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Abstract: *The challenges of physical measurement for prosthetic development and highlight how our integrated tool offers a remote solution to address these challenges, promoting accessibility, convenience, and active participation in the prosthetic design process. Our Measurement Module for Prosthetic Hand Dimensions is a comparison of Image Processing and Distance Measurement Methods The creation of a module to precisely measure finger dimensions for the creation of prosthetic hands is the main goal of this research project. The module provides distance measurement and image-based input as two different ways to get dimensions. Users place their hands at predetermined distances from a screen in the distance measurement option, while users submit photographs of their hands with reference objects for scale in the image based input option. In order To provide a proper fit and functionality, the collected dimensions are an essential component of the prosthetic hand construction process. The study contrasts the efficiency of distance measurement methods employing cameras or other cameras with image processing techniques like edge recognition and contour analysis. Accuracy, usability, and efficiency are among the factors that are assessed to identify the best method for acquiring dimensions. OpenCV for image processing, TensorFlow for machine learning-based analysis, Tkinter for UI design, Mediapipe for landmark identification and hand tracking, and NumPy for numerical operations are just a few of the open-source libraries that are used in this module. The module's possible influence on improving the creation and customization of prosthetic hands is also covered. By offering insights into effective and precise dimension measurement methodologies for customized prosthetic hand design, the study's findings advance the area of prosthetics.*

Keywords: Image processing, Prosthetic development, OpenCV, TensorFlow, Mediapipe, NumPy, Precision dimension measurement, Tkinter

I. INTRODUCTION

There are several obstacles in the way of designing prosthetic hands that properly suit the special requirements of people with limb variations, especially when it comes to measuring fingers accurately. These metrics have a direct impact on the quality of life of users by guaranteeing that prosthetic devices fit, work, and are comfortable. However, the use of specialised equipment and accessibility issues are just two of the drawbacks associated with traditional physical assessment techniques.[1] Our research project offers a novel solution to these problems in the form of the "Measurement Module for Prosthetic Hand Dimensions." This module offers a comprehensive tool for the precise and remote measurement of finger measurements by integrating cutting-edge technology in image processing, hand tracking, and user interface design. Our project consists of multiple major parts and uses open-source libraries and state-of-the-art technology to accomplish its goals. In the first stage, we carry out a comprehensive requirement analysis, working together with prosthetists and people who have different limbs to comprehend their unique requirements and design preferences. Our measurement module is developed using a variety of libraries and frameworks that we have carefully chosen and integrated. Notable examples of this include Mediapipe, which offers sophisticated hand tracking algorithms for real-time landmark localization and identification, and OpenCV, which is used for image processing tasks like hand recognition and feature extraction. Tkinter is selected for UI design to offer a user-friendly interface with straightforward controls and visualisation of measurement data, while TensorFlow is used for machine learning-based

analysis to improve the accuracy and resilience of our dimension measurement techniques. NumPy is also utilised for data manipulation and numerical operations, which makes our module's calculation and analysis more effective.

During the validation and testing stage, we carry out comprehensive trials to evaluate our integrated tool's correctness, dependability, and user happiness. This entails using artificial intelligence (AI), controlled studies, and user trials to verify the module's efficacy and usability in practical prosthetic design situations. By developing a precise and useful dimension measuring instrument that is suited to prosthetic design requirements, our research project hopes to further the area of prosthetics. Our initiative improves the design and customisation of prosthetic hands by utilising cutting-edge technologies and techniques, which in turn enhances the quality of life for people with limb differences. We developed our module by working with prosthetists and people who have different limbs to identify particular needs and preferences for prosthetic design. A few essential elements include MediaPipe for hand tracking, TensorFlow for machine learning-based analysis, Tkinter for user interface design, NumPy for numerical calculations, and OpenCV for image processing. With an emphasis on real-world prosthetic design scenarios, extensive validation and testing guarantee the integrated tool's correctness, dependability, and user happiness.[3] Beyond the near-term obstacles, our initiative hopes to increase accessibility and diversity in prosthetic design while showcasing the revolutionary potential of cutting-edge technologies to transform customised prosthetic solutions. Through the promotion of stakeholder engagement and technological innovation, our project enhances the independence and quality of life for people with limb differences.

II. LITERATURE SURVEY

A number of important topics are covered in the literature review for the Prosthetic Hand Dimension Measurement Module. It begins with an overview of recent studies in the fields of rehabilitation engineering and prosthetics, with a particular emphasis on customised prosthetics and the significance of precise dimension measurements for fitting. Measurement techniques for dimensions, both conventional and digital, are discussed along with research contrasting their accuracy. In addition, studies on distance measurement technologies are reviewed, as well as image processing and computer vision methods for hand analysis. The applicability of popular software libraries and tools like MediaPipe, TensorFlow, and OpenCV is examined. Prosthetics case studies and real-world applications offer valuable information for developing new modules. All things considered, the literature review offers insightful information and practical approaches to efficiently progress the project.

The goal of the paper as presented in the content provided is to incorporate image processing methods into a computer science curriculum while keeping photography fundamentals in mind. It recognises that photography is an interdisciplinary field and makes links between computer science, physics, and the social sciences. Through the provision of both practical laboratories on image processing using open-source tools such as GIMP and ImageJ, and lectures on the principles of photography, students acquire the ability to manipulate images while comprehending the underlying processes. Defining Features Matrix and One Sentence Summary are two examples of classroom assessment tools that are used to measure student understanding and participation. Students employ newly learnt techniques to enhance photos and exhibit a portfolio of their work as part of the course's final project. [1]

With a special emphasis on hand gesture identification, the paper offers a thorough method of gesture recognition using machine learning algorithms and image processing techniques. It highlights the differences between computer analysis of images and human perception when discussing the difficulties in real-time gesture detection. The suggested system uses a support vector machine (SVM) for picture preprocessing, segmentation, contour extraction, and classification. The technology has certain problems, such as image size reduction and feature extraction, but it also offers benefits, such as enhanced human-computer interaction and possible applications in automation and touchless interfaces. All things considered, the study emphasises how versatile gesture recognition systems may be and how crucial it is that more research and development be done in this area. [2]

The creation of a real-time object measurement application using computer vision techniques is covered in the study. The steps in the procedure are taking a picture, identifying the object, setting up coordinates around it, and figuring out how big it is. The system performs mathematical operations and image processing using libraries such as OpenCV and NumPy. The technology successfully estimates the dimensions of different items through simulations, proving its usefulness. The suggested approach makes use of advances in computer vision and artificial intelligence to provide a

practical and precise solution for dimension measurement across a range of sectors. The study also offers insights into its accuracy and performance by comparing its findings with theoretical values. All things considered, the work offers a productive method for measuring an object's dimension in real time that has a wide range of potential uses. [3]

Four well-known Python GUI libraries are summarised in the paper: ipywidgets, Flexx, PySimpleGUI, and Tkinter. Tkinter is a popular choice for GUI programming because of its simplicity, portability, and ease of learning. It has been a part of the Python standard library since 1994. Flexx makes use of web technologies to produce GUIs, making it simple to create GUIs using only Python. In order to illustrate each library's ability to create graphical user interfaces (GUIs) for scientific analysis, system response analysis, and visualisation, the article provides basic examples for each library. Every library has its advantages and is appropriate for a variety of use cases; the selection process is frequently influenced by personal taste. Additional GUI frameworks could be investigated in future studies, and their performance and adaptability could be compared. [4]

The MediaPipe framework connects and abstracts individual perception models into manageable pipelines, making it easier to create augmented reality (AR) applications. It makes it easier to include extra processing stages or inference models by enabling developers to create pipelines as directed graphs of modular components. MediaPipe uses custom calculators to define pipelines and streams for data flow, enabling a range of operations including object detection, face landmark detection, and segmentation. The framework gives developers the ability to optimise and analyse the behaviour of their pipelines by offering tools for performance evaluation and visualisation. Through the use of reusable components and an extensive configuration language, developers can effectively construct and implement intricate perception pipelines, guaranteeing seamless and effective processing of sensory data across various devices. [5]

In specifically, real-time object detection and measurement using machine learning approaches is explored in this research article. It starts with a review of the literature, mentioning important works like the object identification technique YOLO and the real-time uses of LiDAR and OpenCV sensors. During the discussion of real-time object identification techniques, accuracy and speed in dynamic situations are emphasised. The study describes a comprehensive algorithm that combines OpenCV, LiDAR sensors, cameras, and machine learning techniques to recognise and measure object sizes. In order to offer accurate measurements of things, it describes procedures for gathering data, training models, and putting them into practice in real time. Recognising limitations such as sensor precision and data accessibility, it is concluded that the suggested method provides a complete answer for real-time object measurement. [6]

Although it provided high-level data structures, the Python programming language did not have effective support for numerical calculation until the mid-1990s when NumPy was developed. Multidimensional arrays are made possible by NumPy, allowing for high-performance numerical calculations. These arrays provide effective memory management and are distinguished by their strides, data type, and form. In comparison to conventional for loops, NumPy's support for vectorized operations greatly increases computing speed. By extending similar processes to arrays with varying forms, broadcasting increases efficiency even more. In order to handle complicated data types, NumPy additionally offers structured arrays and memory mapping for efficient input/output. Adopted in academia, industry, and scientific research, NumPy's wide range of functionalities and effective design have made it a mainstay in numerical computation. [7]

With an emphasis on Python's use in image processing, the article explores the language's features and wide range of applications. Python may be used for a variety of tasks, from scientific computing to web development, because of its simplicity and versatility. The Python image Library (PIL), which provides functions for picture editing, filtering, and enhancement, is emphasised as a potent tool in image processing. The article examines how Python can be used for tasks like image filtering, contour outlining, and geometric transformations when paired with libraries like PIL and OpenCV. Additionally, because of its abundance of libraries and ease of mastering, the research recommends Python as the best language for experimental programming in image processing. Overall, the article emphasises how Python has advanced digital image processing and advocates for ongoing exploration and development in this field. [8]

III. METHODOLOGY

Creating a module that detects hand and fetch dimension of fingers and store it in csv files. Here is a method for creating the module:

cv2 (OpenCV) Handling images and videos, finding objects, and getting key features is easy with OpenCV's awesome computer vision tools. NumPy (np) NumPy is a math whiz for Python, making multi-dimensional arrays a breeze and crunching numbers like a pro. pandas (pd) Pandas rules the data kingdom, transforming messy CSV files into tidy Data Frames for smooth data wrangling. cv2.VideoCapture(): This opens your computer's camera. . It makes a VideoCapture object. Then, you can take pictures from the camera.

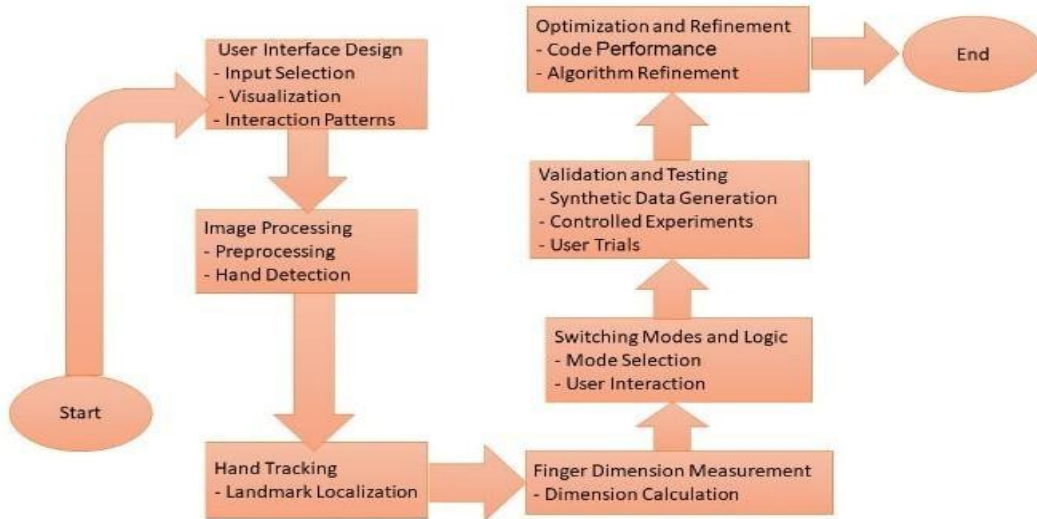


Fig 1: Working of hand detection Module

And do more things with those pictures. Mediapipe Hands Module Google's Mediapipe is a framework for perception tasks using machine learning pipe-lines. Its Hands module detects and tracks hands accurately in video and images. Straight-Line Length Euclidean distance determines the straight path between two points. It finds the hand's position from the camera centre. The formula is simple $\sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$. This calculates the space between the hand and a set camera location. User Feedback, the system instructs the user to modify the location of their hand in relation to the camera based on the determined distance. By doing this, you may be confident that the hand is at the ideal distance for precise measurement. cv2.VideoCapture.read() This method reads a frame from the camera. It returns a Boolean value indicating whether the frame was successfully captured and the frame itself. Next is Feature Extraction This involves using image tricks to pull out important details, like finger size. Techniques may include segmenting the image, finding edges, analysing shapes whatever fits the job. Pandas DataFrame, the hand sizing data is kept in a CSV file using Pandas. This makes it simple to edit, analyze, and share the info. CAD Software or 3D Modelling, the stored hand measurements are used as input for designing a prosthetic digit via CAD software or 3D modelling libraries. This allows creating custom fitted prosthetics tailored to the individual's precise hand dimensions. Grasping the underlying theory helps implement each step efficiently in the code and build a sturdy system for hand measurement and prosthetic design. You can use the Tkinter package in Python to develop a user interface (UI) for your prosthetic design and hand measurement module.

Requirement Analysis:

Perform a comprehensive requirements analysis, taking into account the particular requirements of prosthetic design. Work along with prosthetists and people who have different limbs to determine the most important dimensions for fingers, including length, width, and curve.

Library Selection

Examine several frameworks and libraries that are appropriate for hand tracking, image processing, and developing user interfaces, keeping in mind their compatibility with hardware and software for prosthetic design. Libraries with

strong hand tracking features and real-time processing support should be given priority in order to guarantee a smooth integration into prosthetic design processes

User Interface Design:

Create an interface that is easy to use and meets the requirements of prosthetic designers and technicians. Include functions like standard accessibility standards support, easily navigable controls, and clear visualisation of measurement data. Prosthetic design procedures can be made easier to use and more efficient by optimising interface layout and interaction patterns with consideration for ergonomic principles and user feedback.

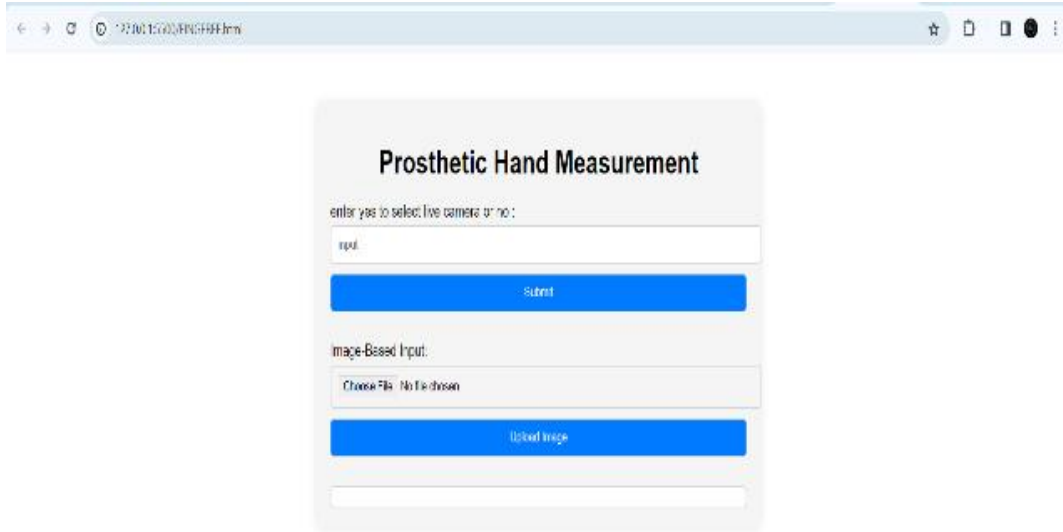


Fig 2 : User Interface

Image Processing:

Create image processing algorithms that preprocess input photos and improve hand detection and landmark localization accuracy by tackling issues including noise, lighting fluctuations, and background clutter. Utilise methods like colour segmentation, edge detection, and morphological procedures to reduce interference from non-hand objects in the image and extract pertinent hand information.

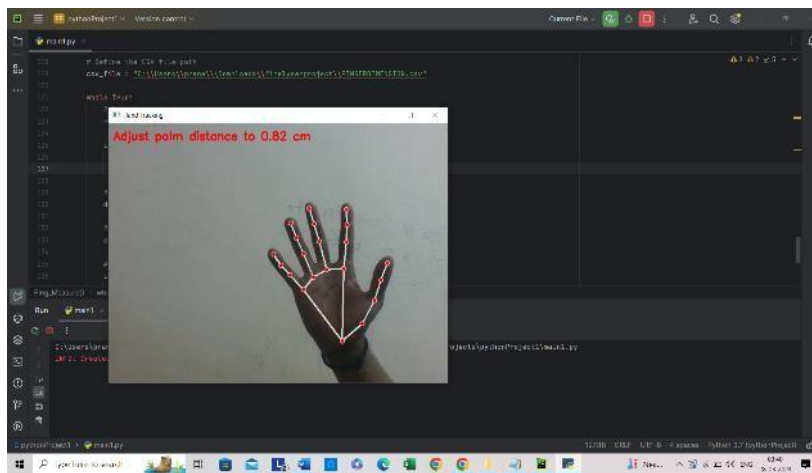


Fig 3: Palm Adjustment

Hand Tracking

Incorporate sophisticated algorithms for hand tracking, like MediaPipe Hands, to precisely identify and monitor hand landmarks in real-time. Hand tracking parameters can be adjusted to balance tracking accuracy and computational efficiency for prosthetic design

Finger Dimension Measurement

Create algorithms that compute the dimensions of fingers by using landmarks on the hand that have been recognised, taking into consideration anatomical variances, perspective distortion, and hand orientation. Regression models and neural networks are examples of machine learning approaches that can be used to improve dimension estimation and adjust to the various hand forms and sizes that are found in prosthetic design.

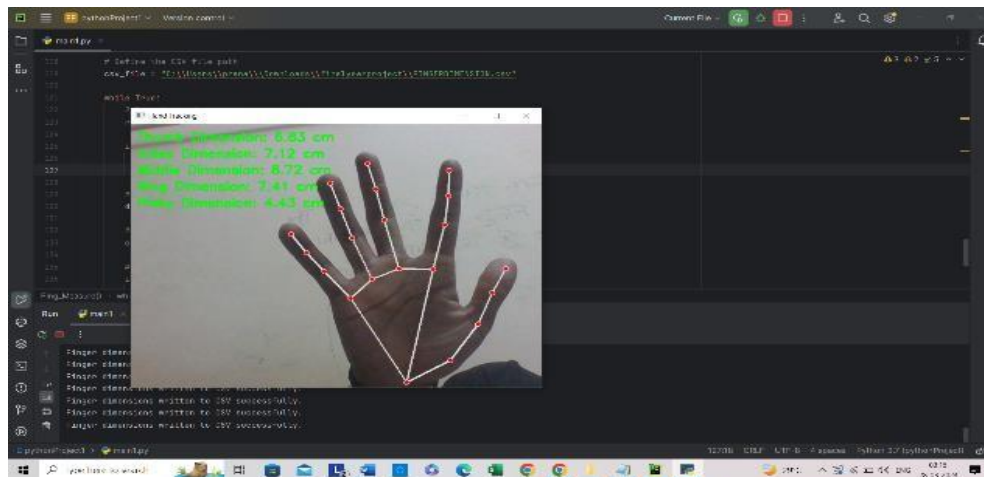


Fig 4 : Fetching Dimension

Switching Modes:

Provide strong logic to handle user input and preserve workflow continuity when alternating between image processing and live camera modes. To reduce confusion and improve usability, give users clear feedback and instructions on how to move between modes and interact with various elements of the integrated product.

IV. CONCLUSION

By providing a state-of-the-art method for accurately and remotely measuring finger dimensions, "Measurement Module for Prosthetic Hand Dimensions" is transforming the prosthetic design process. With the use of cutting edge technology and a user-friendly interface, our module encourages inclusivity, efficiency, and accessibility for people with limb differences. Our goal is to raise the standard of living for prosthetic users all around the world by utilising creative thinking and encouraging teamwork in prosthetic design processes

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