

INNOVATION. AUTOMATION. ANALYTICS

PROJECT ON

Al Powered Solution for Assisting Visually Impaired Individuals

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About me

am currently pursuing a B.Tech degree in Computer Engineering at Pimpri Chinchwad College of Engineering, Pune. With a strong foundation in mathematics, having been selected for the Indian National Mathematical Olympiad (INMO) in 2020, I have always been deeply passionate about problemsolving and analytical thinking. This love for mathematics naturally led me to explore data science, where I can combine my analytical skills with programming to uncover patterns, insights, and solutions in complex datasets. am particularly drawn to data science because it offers a perfect blend of my interests in statistics, machine learning, and data-driven decision-making. The field's ability to transform raw data into actionable knowledge fascinates me, and I love the challenge of making sense of numbers, finding trends, and predicting outcomes.



About me

Beyond academics, I am also passionate about poetry, novel reading, and acting, all of which help me cultivate a creative and critical mindset. These interests have shaped my approach to data science, allowing me to think outside the box and approach problems with a fresh perspective.

This is my first internship, and I am currently enjoying every moment of it. I am gaining a wealth of knowledge, particularly in exploratory data analysis, Python, and machine learning. It has been an enriching experience, providing me with hands-on learning opportunities and a deeper understanding of the data science field.



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Agenda

- Business Problem and Use case domain understanding(If Required)
- Objective of the Project
- Work Flow
- Work Flow
- a. Environment Setup
- b. Core Functionalities
- c. User Interaction
- d. Output Delivery

Conclusion (Key finding overall)

Your Experience/Challenges working on this project.



Business Problem Statement

impaired in

Visually impaired individuals face significant challenges in navigating their environments, detecting obstacles, and performing tasks such as reading text or identifying objects. The problem lies in creating an intelligent, real-time system to assist visually impaired people in understanding their surroundings, recognizing objects, and receiving actionable insights.



Develop a user-friendly Alpowered tool leveraging computer vision and NLP to provide real-time scene descriptions and assistance for visually impaired individuals. Prioritize accessibility, real-time performance, and privacy while continuously improving based on user feedback and technological advancements. Promote the system through collaborations with accessibility-focused organizations.



To empower visually impaired individuals by providing real-time, Al-driven assistance for understanding their surroundings, improving accessibility, independence, and quality of life through advanced technology.

Objective

The primary objective of this project is to develop a robust AI-based vision assistance system that can:

- 1. Detect and highlight objects in images or videos.
- 2. Extract text from images.
- 3. Provide detailed descriptions of scenes and objects.
- 4. Offer voice-based assistance for a hands-free user experience.
- 5. Assist in daily tasks by analyzing the environment and identifying potential obstacles or objects. This application aims to enhance the independence and quality of life of visually impaired users.

Workflow:-





Overview of Workflow

The application consists of the following steps:

1. Environment Setup

- •Import necessary libraries and modules such as streamlit, torch, pytesseract, cv2, and google.generativeai.
- •Configure API keys using dotenv to ensure secure access to external services.
- •Initialize pre-trained models like fasterrcnn_resnet50_fpn for object detection.

2. Core Functionalities

- •Text Extraction from Images: Use pytesseract to extract readable text.
- •Object Detection: Use a pre-trained Faster R-CNN model for detecting objects in images and drawing bounding boxes.
- •Scene Description: Leverage a generative AI model to analyze and describe the scene.
- •**Text-to-Speech Conversion**: Use pyttsx3 to convert extracted text or scene descriptions into speech for auditory feedback.

3. User Interaction

- •Users can upload images or initiate video capture.
- •The app processes the input, performs specified tasks, and returns results in the form of annotated images, text, or speech.

4. Output Delivery

- •Annotated images are displayed with detected objects and their labels.
- •Descriptions are provided for scenes, objects, and text content.
- •Voice-based assistance delivers the outputs audibly.



Environment Setup

1. Import Libraries and Modules

The libraries and modules used in the project are essential for various functionalities, including object detection, text recognition, image processing, and natural language processing.

2. Configuring API Keys with dotenv

Environment variables are used to store sensitive information like API keys securely.

How doteny Works:

Place your keys in a .env file:

The load_dotenv() function reads these values and makes them accessible via os.getenv.

```
♦ VisionAssist.py X  ■ Langchain.ipynb

C: > Users > shind > Python_Coding > API > ♥ VisionAssist.py > ...
       import streamlit as st
       import google.generativeai as genai
       from PIL import Image, ImageDraw
       import pytesseract
       import pyttsx3
       import torch
       from torchvision import transforms
       from torchvision.models.detection import fasterrcnn resnet50 fpn
       from dotenv import load dotenv
       import cv2
       import os
       import uuid
       import time
       from io import BytesIO
```

```
VisionAssist.py X
Langchain.ipynb
C: > Users > shind > Python_Coding > API > VisionAssist.py > ...

16  # Loading the environment variables
17  load_dotenv()
18
19  # Loading API key
20  with open(r"C:\Users\shind\Python_Coding\API\keys\API_key.txt") as f:
21  api_key = f.read()
22
23  genai.configure(api_key=api_key)
```



Environment Setup

3. Pre-trained Model Initialization

fasterrcnn_resnet50_fpn is a pre-trained object detection model available in PyTorch's torchvision.

Model Details:

Faster R-CNN: A robust object detection algorithm. **ResNet-50 Backbone:** Used for feature extraction.

Pretrained=True: Loads a model trained on the COCO dataset.

4. Installation Commands

To install the required libraries, use the following commands in your terminal or command prompt:

pip install streamlit

pip install torch torchvision

pip install pytesseract

sudo apt install tesseract-ocr # For Linux systems

brew install tesseract # For macOS

pip install opency-python

pip install Pillow

pip install pyttsx3

pip install python-dotenv

pip install google-generativeai



Environment Setup

5. Setup Tips

Tesseract Path Configuration (Windows):

If Tesseract is not in your PATH, specify its location in your code:

pytesseract.pytesseract.tesseract_cmd = r'C:\Program
Files\Tesseract-OCR\tesseract.exe'

API Key Placement:

Store sensitive keys like API keys in a .env file to avoid hardcoding them into the script.

By setting up these libraries and pre-trained models, the project integrates key functionalities like object detection, text recognition, and Al-driven scene description efficiently.



Core Functionalities

1. Text Extraction

Purpose: Extracts text from an image and narrates it.

Functionality:

Opens an uploaded image and uses pytesseract to extract text. Handles errors gracefully, providing user feedback when text extraction fails.

How it Works:

Uses pytesseract to process the uploaded image and extract text. Text is displayed on the interface and narrated using pyttsx3.

2. Object Detection

Purpose: Identifies objects in the image, highlights them, and narrates key details.

Functionality:

Converts an image to a tensor for processing.

Performs object detection using a Faster R-CNN model.

Applies non-maximum suppression (NMS) to eliminate overlapping detections.

How it Works:

Converts the image into a tensor using torchvision.transforms.

Runs it through a pre-trained Faster R-CNN model

(fasterrcnn_resnet50_fpn) for object detection.

Uses non-maximum suppression to filter overlapping bounding boxes.

Output Highlighting:

Highlights objects on the image with bounding boxes and labels.

```
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C: > Users > shind > Python_Coding > API > 🔷 VisionAssist.py > 🔀 process_video
      def perform_object_detection(image, threshold=0.5, nms_threshold=0.5):
               preprocess = transforms.Compose([transforms.ToTensor()])
               image tensor = preprocess(image)
               outputs - detection model([image tensor])[0]
               indices = torch.ops.torchvision.nms(outputs['boxes'], outputs['scores'], nms_threshold)
               return (k: v[indices] for k, v in outputs.items() if k in ['boxes', 'labels', 'scores']]
          except Exception as e:
               raise RuntimeError(f"Object detection failed: (e)")
      # Draw Detected Objects on Image
       def highlight_objects(image, detections, threshold=0.5):
           for box, label, score in zip(detections['boxes'], detections['labels'], detections['scores'])
                  x1, y1, x2, y2 - box
                  draw.rectangle([x1, y1, x2, y2], outline="red", width=2)
                  class name = COCO CLASSES[label.item()]
                  draw.text((x1, y1), + {class_name} ((score .1f})", fill-"yellow")
          return image
```

Core Functionalities

3. Scene Description

Purpose: Provides an Al-generated description of the environment and objects in an image.

Functionality:

Sends an image to a generative AI model for scene analysis.

Provides descriptive text highlighting objects, people, and activities.

How it Works:

Uses the Gemini model from Google Generative AI to generate descriptions.

Designed to be concise, focusing on essential elements like objects, people, and actions.

4. Text-to-Speech Conversion

Purpose: Converts generated text into speech for auditory feedback.

Functionality:

Uses pyttsx3 to convert text into speech.

Maintains session state to ensure efficient audio playback.

How it Works:

Initializes a text-to-speech engine using pyttsx3.

Plays audio feedback of the generated descriptions or detected text.

```
VisionAssist.py
■ Langchain.ipynb
C: > Users > shind > Python_Coding > API > ♥ VisionAssist.py > ♥ describe_scene
       def describe_scene(image_path):
           """Describe the scene in the given image using an AI model."""
               img = Image.open(image path)
               prompt = (
                        "Analyze the uploaded image and describe the scene and object in clear and"
                       "simple language in very short to assist visually impaired users like"
                       "crossing roads and detecting objects. Answer must include key"
                       "details with highlights about the environment, objects, people, and"
119
                        actions present in the image."
               # Assuming genai is pre-configured with your API key
               response = genai.GenerativeModel("gemini-1.5-pro").generate_content([prompt, img]).text
               return response.strip()
           except Exception as e:
               return f"Error describing scene: {e}"
```



Core Functionalities

5. Real-Time Video Processing

Purpose: Captures live video, extracts frames, and processes them for object detection and scene description.

Functionality:

Captures video from the webcam, processes frames, and describes each frame's content.

Stops video capture after a set time to ensure timely output.

How it Works:

Captures frames using OpenCV.

Saves each frame and processes it for scene descriptions and object detection.

Narrates the scene description for user assistance.

```
♦ VisionAssist.py

■ Langchain.ipynb
C; > Users > shind > Python_Coding > API > ♥ VisionAssist.py > ♥ describe_scene
      def process video():
           """Capture video, save frames, and describe scenes."""
           cap = cv2.VideoCapture(0) # 0 for default webcam
           frame count = 0
           start time = time.time() # Start timer when capture begins
           pause time = 15 # Time (in seconds) to capture video before pausing
           while cap.isOpened():
               ret, frame = cap.read()
               if not ret:
                   break
               # Calculate elapsed time
               elapsed time = time.time() - start time
               # Save the current frame
               frame path = save frame(frame, frame count)
               # Describe the scene (always process each frame)
               scene description = describe scene(frame path)
               # Display the processed frame and description in Streamlit
               st.image(frame, channels="BGR", caption=f"Frame {frame count}")
               st.write(f"Frame {frame count} Description: {scene description}")
               speak text(scene description)
               frame count += 1
               if elapsed_time >= pause_time: # Pause after 15 seconds
                   st.write("Video capture paused after 15 seconds.")
                   break # Exit the loop after 15 seconds of capture
           cap.release()
           st.success("Video processing completed. All frames saved.")
```



User Interaction

User Interaction

This section outlines how users interact with the application. Users can upload images or capture videos, and the app processes the input to return results in various formats, such as annotated images, extracted text, or synthesized speech.

User Interaction Flow

Image Upload: Users upload an image through the app interface.

Video Capture: Users can initiate video capture using their webcam.

Processing: The app performs tasks such as object detection, text recognition, or scene description.

```
VisionAssist.py
Langchain.ipynb
C: > Users > shind > Python Coding > API > VisionAssist.py > ...
      # Streamlit interface
      st.title("Real Time Video Assist for Visually Impaired")
      if st.button("Start Video Capture"):
          process video()
      # Sidebar with Buttons
      st.sidebar.title("Actions to Perform")
      text_button = st.sidebar.button("Describe Image ] ")
      scene button = st.sidebar.button("Read Text from Image /")
      detect button = st.sidebar.button("Detect Objects ! ")
      assist_button = st.sidebar.button("Personal Assist")
      audio_button = st.sidebar.button("Stop Audio ♥")
      # Center Layout for Upload Section
180
      st.title("Image Assistance for Visually Impaired")
      uploaded_images = st.file_uploader("Upload multiple images:", type=["jpg", "jpeg", "png"], accept_multiple_files=True)
```



1. Text Extraction (if text_button)

This section handles the generation of a textual description of the uploaded image.

Functionality:

Prompts the Generative AI model to analyze the uploaded image and provide a clear description of the scene for visually impaired users.

Focuses on generating text that highlights the environment, objects, people, and actions within the image.

Process:

Converts the uploaded image into bytes for compatibility with the Al model.

Passes a descriptive prompt along with the image to the generative model.

Displays the Al-generated description to the user.

Output:

A written description of the image scene.

```
VisionAssist.py
Langchain.ipynb
C: > Users > shind > Python_Coding > API > ♥ VisionAssist.py > ...
       for uploaded image in uploaded images:
           # Process each image
          st.image(uploaded_image, caption=f"Uploaded Image: {uploaded_image.name}", use_column_width=True)
           # Text Extraction
          if text button:
               with st.spinner("Describing image..."):
                   prompt = (
192
                       "Analyze the uploaded image and describe the scene in clear and"
                       "simple language to assist visually impaired users. Include key"
                       "details about the environment, objects, people, and actions present in the image."
                   image bytes = convert image to bytes(uploaded image)
                   response = genai.GenerativeModel("gemini-1.5-pro").generate content([prompt, image bytes[0]]).text
                   st.subheader("Image Description:")
                   st.write(response)
```



2. Scene Description (if scene_button)

This section is similar to text extraction but focuses more on detailed scene analysis.

Functionality:

Provides an in-depth explanation of the scene, identifying specific objects, people, and their interactions. Highlights key environmental details in clear language.

Process:

Converts the uploaded image into bytes.

Sends a scene-specific prompt to the AI model.

Displays the description and optionally speaks the text for the user.

Output:

A detailed scene description emphasizing all objects and environmental features.



3. Object Detection (if detect_button)

This section performs object detection on the uploaded image.

Functionality:

Detects objects in the image and annotates it with bounding boxes and labels.

Enhances the image to provide visual feedback on identified objects.

Process:

Loads the uploaded image.

Uses a function (perform_object_detection) to detect objects.
Annotates the image using a helper function (highlight_objects).
Displays the annotated image in the app.

Output:

An annotated image showing detected objects and their locations.

```
VisionAssist.py ● Langchain.ipynb

C: > Users > shind > Python_Coding > API > ◆ VisionAssist.py > ...

215  # Object Detection
216  if detect_button:
217  with st.spinner("Detecting objects..."):
218  img = Image.open(uploaded_image)
219  detections = perform_object_detection(img)
220  annotated_image = highlight_objects(img.copy(), detections)
221  st.image(annotated_image, caption="Objects Detected", use_column_width=True)
```



4. Assistance (if assist_button)

This section provides tailored assistance for visually impaired users based on the uploaded image.

Functionality:

Acts as a virtual assistant to analyze the image and provide useful insights.

Focuses on helping users with real-world tasks, such as identifying obstacles, recognizing objects, or describing the surrounding environment.

Process:

Converts the uploaded image into bytes.

Sends a detailed assistance prompt to the generative model.

Displays the Al-generated response and optionally speaks it for the user.

Output:

A helpful description or analysis aimed at assisting visually impaired users.

```
VisionAssist.pv
Itanochain.ipvnb
C: > Users > shind > Python_Coding > API > ♥ VisionAssist.py > ...
           # Assistance
           if assist button:
               with st.spinner("Providing assistance..."):
                   assistance prompt =
                       "You are a helpful AI assistant designed for Visually impaired people."
                      "Analyze the uploaded image (Boxes represent any obstacles or objects)"
                       "and identify obstacles (and give numerically distance from obstacles,"
230
                       "if car give real name) or objects so you can assist them with, tasks like"
                       "crossing roads, playing with animals, doing their daily tasks, telling environment"
                       "around them , recognizing objects or reading labels."
                   image bytes = convert image to bytes(uploaded image)
                   assistance response = genai.GenerativeModel("genini-1.5-pro").generate content([assistance prompt, image bytes[0]]).text
                  st.subheader("Assistance:")
                   st.write(assistance response)
                   speak text(assistance response)
```



5. Audio Playback Control (if audio_button)

This section handles stopping text-to-speech (TTS) playback.

Functionality:

Allows users to stop the playback of any previously generated audio.

Process:

Initializes a text-to-speech (TTS) engine if it hasn't been initialized yet.

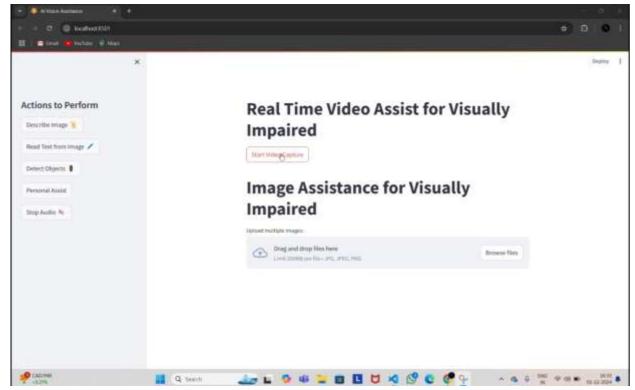
Stops any ongoing audio playback using the TTS engine. Displays a success or error message depending on the outcome.

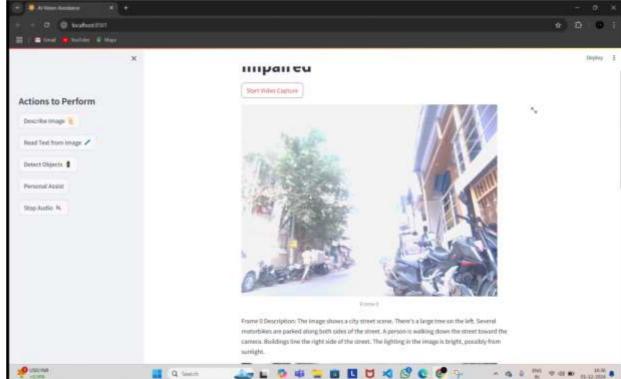
Output:

Stops any active audio playback and informs the user.

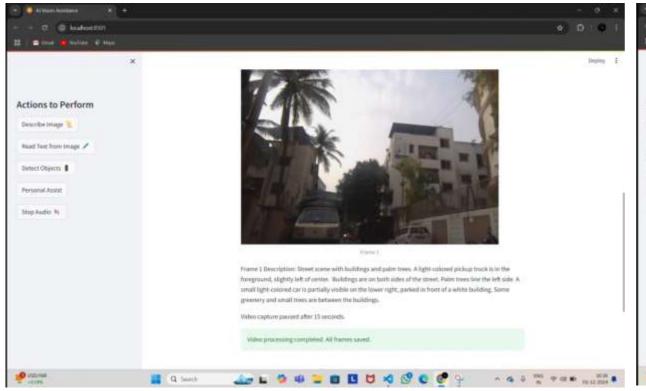
```
Langchain.ipynb
VisionAssist.py
C: > Users > shind > Python_Coding > API > 🕏 VisionAssist.py > ...
241
       if audio button:
242
           try:
               # Initialize TTS engine if not already initialized
               if "tts engine" not in st.session state:
                   st.session state.tts engine = pyttsx3.init()
247
               # Stoping the audio playback
               st.session state.tts engine.stop()
               st.success("Audio playback stopped.")
250
           except Exception as e:
251
               st.error(f"Failed to stop the audio. Error: {e}")
```

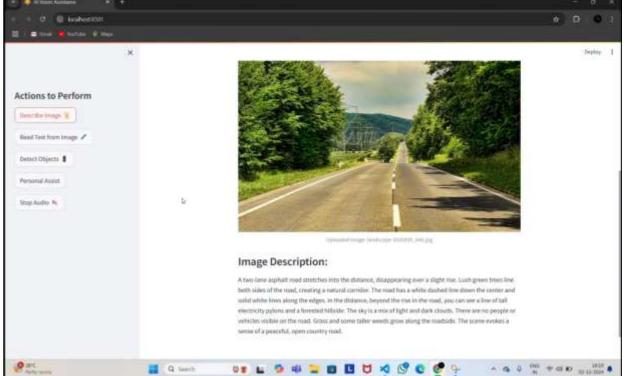




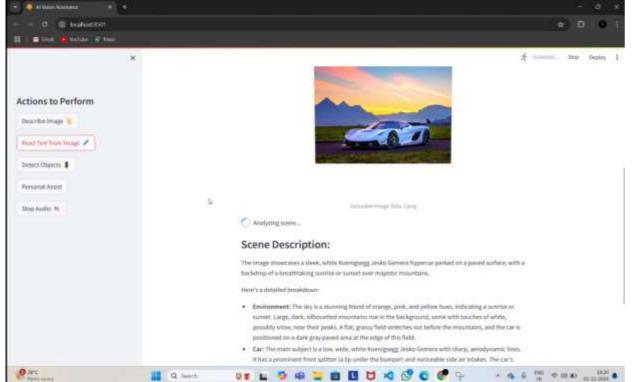


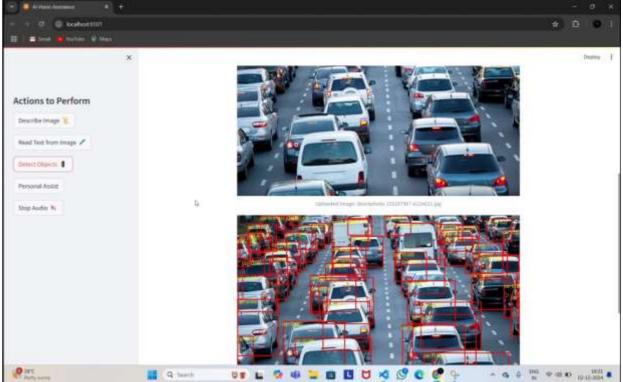




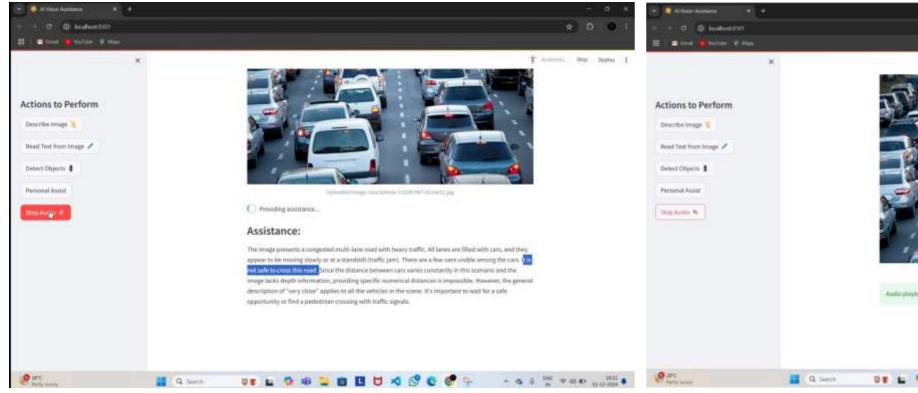


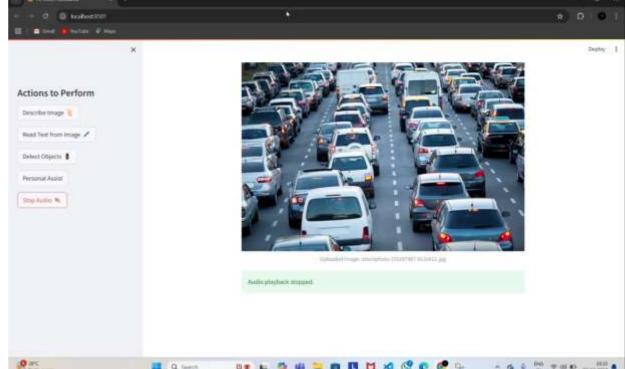














Final Conclusion

This application integrates advanced computer vision, AI, and TTS technologies to offer a comprehensive vision assistance tool. Each functionality enhances the independence of visually impaired users by enabling them to:

- 1. Read text in images.
- 2. Identify and locate objects in real-time.
- 3. Understand and interact with their environment effectively.

Future improvements can include expanding object detection capabilities, enhancing scene descriptions with depth information, and integrating multilingual support. Additionally, leveraging edge computing could improve response times and allow offline functionality, ensuring uninterrupted usability. Incorporating gesture recognition and voice-command interfaces could further simplify user interaction, providing a seamless experience.



Experience:

1. Knowledge of Core Libraries

l explored various powerful libraries:

Image Processing:

Learned to use OpenCV or Pillow for tasks like image resizing, filtering, and feature extraction.

Gained expertise in image enhancement techniques (e.g., denoising, edge detection).

Machine Learning:

Worked with TensorFlow, PyTorch, or scikit-learn for training or deploying models.

Understood model pipelines, including loading pre-trained models.

OCR (Optical Character Recognition):

Used libraries like Tesseract or EasyOCR to extract text from images.

Improved understanding of handling text recognition in noisy or low-light images.

2. Working with APIs and Pre-trained Models

Integrated APIs like Google Vision API.

Understood the importance of transfer learning for tasks like object detection, text recognition, or scene analysis.

Learned to balance between using lightweight models for speed and heavier models for accuracy.

3. Real-time Processing

Built pipelines for real-time image capture using camera feeds and webcams.

Mastered threading or asynchronous techniques to manage the processing speed of vision tasks.

4. Data Handling and Annotation

Gained insights into preparing datasets for training and testing:

Labeling images for object detection or classification tasks.

Understanding data augmentation techniques to improve model generalization.

5. Debugging and Performance Tuning

Handled issues with model inference speed, memory usage, and accuracy trade-offs.

Debugged cases where the system failed in recognizing or detecting objects accurately.

Overall Takeaway

This project taught you not just technical skills but also problem-solving, adaptability, and innovation. It's a journey that highlights the blend of software engineering, AI, and user-focused design, making you more proficient in cutting-edge technologies and their practical applications.

THANK YOU



