Automating Lateral Flow Tests This project automates the analysis of lateral flow test (LFT) images to determine results ("Positive", "Negative", or "Void") with confidence scores. The pipeline includes quality checks, background removal, red channel highlighting, cropping, and line detection. #importing libraries required import os import cv2 import numpy as np from rembg import remove #deep learning based background removal from PIL import Image from pathlib import Path import matplotlib.pyplot as plt from collections import Counter from tqdm import tqdm Step 1: Image Quality Assessment and Background Removal Images are assessed for brightness and sharpness to ensure reliable LFT analysis. Poor-quality images (too dim or blurry) are rejected, and backgrounds are removed to isolate the test strip. • Brightness: Normalized histogram average (0-1). • Sharpness: Laplacian variance (higher = sharper). • Cropping: Uses largest contour to focus on LFT strip. In []: #function to calculate brightness (using average pixel intensity) #calculates the normalized average pixel intensity of the image to assess its brightness. def calculate brightness(image): grayscale = image.convert("L") #Converts image to grayscale to simplify brightness calculation. histogram = grayscale.histogram() pixels = sum(histogram) brightness = sum(i * pixel for i, pixel in enumerate(histogram)) / pixels return brightness / 255 #normalize to [0, 1] #function to calculate sharpness (using laplacian variance) #calculates the variance of the Laplacian, which measures the sharpness of edges in the image def calculate sharpness(image): #measures edge sharpness using laplacian variance image np = np.array(image.convert("L")) laplacian_var = cv2.Laplacian(image_np, cv2.CV_64F).var() return laplacian var #higher values indicate sharper images #function to crop the image to remove black background #crops the image by isolating the largest contour, typically representing the region of interest (lft strip) def crop to lft(image): #isolates largest contour (lft strip) for cropping image np = np.array(image) gray = cv2.cvtColor(image np, cv2.COLOR RGB2GRAY) #convert to grayscale for contour detection. , binary = cv2.threshold(gray, 1, 255, cv2.THRESH BINARY)#create a binary image for contour finding. contours, _ = cv2.findContours(binary, cv2.RETR_EXTERNAL, cv2.CHAIN_APPROX_SIMPLE) #find contours in the binary image. largest contour = max(contours, key=cv2.contourArea) #select the largest contour, assumed to be the lft region. x, y, w, h = cv2.boundingRect(largest contour) cropped image = image.crop((x, y, x + w, y + h)) return cropped image #returns focused lft region #paths to folders input folder = "/Users/saniyakhan/Desktop/Option 1 data" output_folder = "/Users/saniyakhan/Desktop/Option 1 data_no_bg" processed folder = "/Users/saniyakhan/Desktop/Processed images" #create output folders if they don't exist if not os.path.exists(output folder): os.makedirs(output folder) #for background removed images if not os.path.exists(processed_folder): os.makedirs(processed folder) #for quality checked images #brightness and sharpness thresholds brightness threshold = 0.1 sharpness_threshold = 5 #lists to track rejected images rejected images = [] #loop through all images in the input folder for filename in os.listdir(input folder): if filename.lower().endswith(('.png', '.jpg', '.jpeg', '.bmp', '.gif')): #full file path input path = os.path.join(input folder, filename) output_path = os.path.join(output_folder, filename) processed_path = os.path.join(processed_folder, filename) #open the image with open(input path, 'rb') as i: #remove the background using rembg input image = i.read() output_image = remove(input_image) #save the output image with open(output_path, 'wb') as o: o.write(output_image) #open the background removed image for further processing image = Image.open(output_path) #convert rgba to rgb if necessary if image.mode == 'RGBA': image = image.convert('RGB') #ensures consistent format #crop the image to remove black background image = crop to lft(image) #check brightness brightness = calculate brightness(image) if brightness < brightness threshold:</pre> print(f"Rejected {filename}: Too dim (Brightness: {brightness:.2f})") rejected images.append((filename, f"Too dim (Brightness: {brightness:.2f})")) continue #check sharpness sharpness = calculate sharpness(image) if sharpness < sharpness threshold:</pre> print(f"Rejected {filename}: Too blurry (Sharpness: {sharpness:.2f})") rejected images.append((filename, f"Too blurry (Sharpness: {sharpness:.2f})")) continue #save the processed image image.save(processed path) print(f"Accepted {filename} (Brightness: {brightness:.2f}, Sharpness: {sharpness:.2f})") #print rejected images at the end if rejected images: print("\nRejected Images:") print("=" * 30) for filename, reason in rejected_images: print(f"{filename}: {reason} - Please retake image.") else: print("\nAll images passed the checks. No rejected images.") print("\nProcessing complete!") **Step 2: Highlighting Red Channels** LFT lines are red, so this step isolates them using HSV color space, which is more robust than RGB for color detection. Why HSV: Separates hue from lighting, helping us improving accuracy. • Range: Tuned to LFT red hues (127-179). In []: #function to highlight red channels in an image using HSV color segmentation. #the processed image is saved in the specified output folder. def highlight red channels(image path, output folder): #load the image image = cv2.imread(image path) if image is None: print(f"Error: Could not load image {image path}") return #convert to HSV for better red color segmentation hsv = cv2.cvtColor(image, cv2.COLOR BGR2HSV) #define the range for red color in HSV (tuned for lateral flow test lines) lower_red = np.array([127, 63, 0]) #lower bounds for red hue upper_red = np.array([179, 255, 255]) #uupper bounds for red hue #create a binary mask for red hues within the defined range red mask = cv2.inRange(hsv, lower red, upper red) #apply the mask to retain only the red regions in the image red highlighted = cv2.bitwise and(image, image, mask=red mask) #save the image with highlighted red regions to the output folder output path = os.path.join(output folder, os.path.basename(image path)) cv2.imwrite(output_path, red_highlighted) print(f"Highlighted red channels for {os.path.basename(image_path)}") #path to the folder of processed images processed folder = "/Users/saniyakhan/Desktop/Processed images" #create the output folder for highlighted images (if it doesn't exist) output folder = "/Users/saniyakhan/Desktop/Highlighted images" if not os.path.exists(output folder): os.makedirs(output folder) #ensure the folder exists #loop through all processed images in the folder and highlight red channels for filename in os.listdir(processed folder): if filename.lower().endswith(('.png', '.jpg', '.jpeg', '.bmp', '.gif')): image path = os.path.join(processed folder, filename) highlight_red_channels(image_path, output_folder) print("Red channel highlighting complete!") Step 3: Cropping to Line Region Images are cropped to the central region where LFT lines typically appear, reducing noise and focusing analysis. • Parameters: 40% width, 40% height, centered. Helps us prevent errors and false positives. In []: #function to crop images to middle region def crop to middle region(input folder, output folder, crop width percentage=0.4, crop height percentage=0.5, vertical offset percentage=0.0): #create output folder if it doesn't exist os.makedirs(output folder, exist ok=True) #get list of image files image_extensions = ['.jpg', '.jpeg', '.png', '.bmp', '.tif', '.tiff'] image_files = [] for file in os.listdir(input folder): ext = os.path.splitext(file)[1].lower() if ext in image extensions: image files.append(file) print(f"Found {len(image_files)} images in folder") #process each image for image_file in image_files: input path = os.path.join(input folder, image file) output_path = os.path.join(output_folder, f"cropped_{image_file}") #read image image = cv2.imread(input_path) if image is None: print(f"Could not read image: {input path}") continue #get dimensions height, width = image.shape[:2] #calculate horizontal crop boundaries crop width = int(width * crop width percentage) left boundary = (width - crop width) // 2 right_boundary = left_boundary + crop_width #calculate vertical crop boundaries crop height = int(height * crop height percentage) vertical_offset = int(height * vertical_offset_percentage) top_boundary = (height - crop_height) // 2 + vertical_offset bottom boundary = top boundary + crop height # ensure boundaries are within image dimensions top boundary = max(0, top boundary)bottom_boundary = min(height, bottom_boundary) #crop image to middle region cropped image = image[top boundary:bottom boundary, left boundary:right boundary] #save cropped image cv2.imwrite(output path, cropped image) print(f"Cropped and saved: {output_path}") #folder paths input folder = "/Users/saniyakhan/Desktop/Highlighted images" output folder = "/Users/saniyakhan/Desktop/Cropped images" #percentage of middle region to keep crop width percentage = 0.4 #40% of width focuses on lines crop_height_percentage = 0.4 #40% of height captures line region vertical offset percentage = 0 #center vertically #process images crop to middle region(input folder, output folder, crop_width_percentage, crop height percentage, vertical offset percentage) print("Cropping completed!") Step 4: Line Detection and Result Classification This step detects red lines in LFT images and classifies results: • Method: Red channel row sums, smoothed with convolution, peak detection. Noise reduction via smoothing; the minimum distance (40 pixels) between peaks. • Confidence: Based on peak intensity and sharpness, capped at 95%. • Results: "Positive" (2 lines), "Negative" (1 line), "Void" (0 lines). In []: #function to analyze lateral flow test def analyze lateral flow test(image path, min line distance=40): #read the image image = cv2.imread(image path) if image is None: return "Error: Could not read image", 0, None #create visualisation copy visualization = image.copy() #extract the red channel _, _, red_channel = cv2.split(image) #calculate row wise sum of red values row sums = np.sum(red channel, axis=1) #smooth the sums to reduce noise smoothed sums = np.convolve(row sums, np.ones(5)/5, mode='same') #find peaks above average avg sum = np.mean(smoothed sums) threshold = avg_sum * 2 #twice average better accuracy #find potential peaks raw peaks = [] window_size = 10 for i in range(window size, len(smoothed sums) - window size): window = smoothed_sums[i-window_size:i+window_size] if (smoothed sums[i] == max(window) and smoothed sums[i] > threshold): raw_peaks.append((i, smoothed_sums[i])) #sort peaks by intensity raw peaks.sort(key=lambda x: x[1], reverse=True) #filter peaks by minimum distance peaks = [] for pos, intensity in raw peaks: if all(abs(pos - existing_pos) >= min_line_distance for existing_pos in peaks): peaks.append(pos) #ensures distinct lines #sort peaks by position (top to bottom) peaks.sort() #calculate confidence based on intensity and sharpness confidence = 0 peak_intensities = [] peak sharpness = [] #get peak intensities relative to threshold for peak in peaks: rel intensity = smoothed sums[peak] / threshold peak_intensities.append(rel_intensity) #calculate peak sharpness if peak > window_size and peak < len(smoothed_sums) - window_size:</pre> local bg = (smoothed sums[peak-window size] + smoothed sums[peak+window size]) / 2 sharpness = smoothed sums[peak] / max(local bg, 1) #avoids division by zero peak_sharpness.append(sharpness) #determine result and confidence line_count = len(peaks) if line count == 0: result = "Void: No lines detected" confidence = 0 #no peaks = zero confidence elif line count == 1: result = "Negative: Control line only" if peak_intensities and peak_sharpness: intensity factor = min(2.0, peak intensities[0]) / 2.0 sharpness_factor = min(3.0, peak_sharpness[0]) / 3.0 confidence = min(95, 50 * intensity factor + 45 * sharpness factor) elif line count == 2: result = "Positive: Control and test lines" if len(peak intensities) >= 2 and len(peak sharpness) >= 2: intensity1 = min(2.0, peak intensities[0]) / 2.0 sharpness1 = min(3.0, peak_sharpness[0]) / 3.0 intensity2 = min(2.0, peak intensities[1]) / 2.0 sharpness2 = min(3.0, peak sharpness[1]) / 3.0conf1 = 50 * intensity1 + 45 * sharpness1 conf2 = 50 * intensity2 + 45 * sharpness2 confidence = min(95, min(conf1, conf2)) else: result = f"Warning: {line_count} lines detected" confidence = 20 #low confidence for unexpected cases #draw detected lines with arrows on visualisation for i, peak in enumerate(peaks): **if** i == 0: color = (0, 255, 0) #green for control line label = "Control" else: color = (0, 0, 255) #red for test line label = "Test" #draw arrow pointing to the line start point = (10, peak) end_point = (50, peak) cv2.arrowedLine(visualization, start_point, end_point, color, 2, tipLength=0.3) #add label text cv2.putText(visualization, label, (55, peak + 5), cv2.FONT_HERSHEY_SIMPLEX, 0.6, color, 2) #add result text with confidence result text = f"{result}" confidence text = f"Confidence: {confidence:.1f}%" cv2.putText(visualization, result_text, (10, 30), cv2.FONT_HERSHEY_SIMPLEX, 0.7, (0, 0, 255), 2) cv2.putText(visualization, confidence_text, (10, 60), cv2.FONT_HERSHEY_SIMPLEX, 0.7, (0, 0, 255), 2) return result, confidence, visualization #function to process all images def process images(folder path, output folder=None): #check if folder exists if not os.path.exists(folder_path): print(f"Error: Folder {folder_path} does not exist") return #create output folder if specified if output folder and not os.path.exists(output folder): os.makedirs(output_folder) #for result images #get all image files image_extensions = ['.jpg', '.jpeg', '.png', '.bmp', '.tif', '.tiff'] image files = [] for file in os.listdir(folder_path): ext = os.path.splitext(file)[1].lower() if ext in image extensions: image_files.append(file) #sort files for consistent output image_files.sort() #process each image results = {} print("\nLateral Flow Test Results:") print("="*60) print(f"{'Filename':<25} {'Result':<30} {'Confidence'}")</pre> print("="*60) plt.figure(figsize=(10, 8)) for image file in image files: image path = os.path.join(folder_path, image_file) result, confidence, visualization = analyze lateral flow test(image path) #store the results results[image file] = { "result": result, "confidence": confidence #print the result print(f"{image_file:<25} {result:<30} {confidence:.1f}%")</pre> #display image with detected lines plt.clf() plt.imshow(cv2.cvtColor(visualization, cv2.COLOR_BGR2RGB)) plt.title(f"{image_file}: {result} (Confidence: {confidence:.1f}%)") plt.axis('off') plt.tight layout() plt.show() #save result image if output folder is specified if output folder: output_path = os.path.join(output_folder, f"result_{image_file}") cv2.imwrite(output_path, visualization) plt.close() #print summary positives = sum(1 for r in results.values() if "Positive" in r["result"]) negatives = sum(1 for r in results.values() if "Negative" in r["result"]) voids = sum(1 for r in results.values() if "Void" in r["result"]) warnings = sum(1 for r in results.values() if "Warning" in r["result"]) print("\nSummary:") print(f"Total images: {len(results)}") print(f"Positive results: {positives}") print(f"Negative results: {negatives}") print(f"Void results: {voids}") print(f"Warnings: {warnings}") #calculate average confidence by result type if positives > 0: avg_pos_conf = sum(r["confidence"] for r in results.values() if "Positive" in r["result"]) / positives print(f"Average confidence for positive results: {avg pos conf:.1f}%") if negatives > 0: avg_neg_conf = sum(r["confidence"] for r in results.values() if "Negative" in r["result"]) / negatives print(f"Average confidence for negative results: {avg_neg_conf:.1f}%") return results #specify the folder path input_folder = "/Users/saniyakhan/Desktop/Cropped images" #output folder to save final results images output folder = "/Users/saniyakhan/Desktop/Results" #process all images in the folder results = process images(input folder, output folder) Interpretation of Results Using our results, we can produce graphs to visualise the results for better interpretation. Below we have include: 1. Bar graph - to present the outcome in a visual and easy to read way. 2. Scatter plot - Image wise variation of confidence percentages to visualise how accurate the pipeline is. In []: #extracting result types by creating lists classifications = [] confidences = [] filenames = [] for filename, data in tqdm(results.items(), desc="Loading result data"): #extract the classification label ("Positive", "Negative", "Void") from the result string label = data['result'].split(':')[0].strip() confidence = data['confidence'] #extract the confidence score for the image classifications.append(label) confidences.append(confidence) filenames.append(filename) #generate a bar chart to visualize the classification distribution (Positive, Negative, Void, etc.) #'counts' is a Counter object that will count the occurrences of each classification type. plt.figure(figsize=(6, 4)) counts = Counter(classifications) plt.bar(counts.keys(), counts.values(), color=['#6baed6', '#74c476', '#fd8d3c', '#d73027']) plt.title('Classification Results') plt.xlabel('Result Type') plt.ylabel('Number of Images') plt.grid(axis='y', linestyle='--', alpha=0.6) plt.tight_layout() plt.show() #generate a scatter plot to visualize the confidence scores for each image. #this plot helps assess how confident the model is for each image, from 0% to 100%. plt.figure(figsize=(10, 5)) plt.scatter(range(len(confidences)), confidences, color='#2c7fb8') plt.xticks(range(len(filenames)), filenames, rotation=90) plt.ylim(0, 100) plt.title('Confidence Scores per Image') plt.xlabel('Image Filename') plt.ylabel('Confidence (%)') plt.grid(True, linestyle='--', alpha=0.5) plt.tight layout() plt.show() In []: