

Image Preprocessing and Feature Extraction for Face and Gesture Recognition

1. Introduction

Image preprocessing and feature extraction are crucial steps in computer vision tasks, particularly in applications requiring high accuracy, such as face and gesture recognition. This report documents the methodology used for preprocessing and feature extraction from images, including noise reduction, enhancement, edge detection, face detection, gesture area extraction, and skeleton extraction. This approach is geared toward preparing datasets for machine learning models focused on recognition tasks.

2. Methodology

2.1 Image Preprocessing

Image preprocessing was conducted to enhance quality and prepare images for feature extraction. Several key steps were involved:

1. **Gamma Correction:** Gamma correction adjusts the brightness of the image, enhancing details in dark or bright regions.
2. **Denoising:** Gaussian noise was added, followed by mean and median filtering to reduce noise while preserving edges.
3. **Edge Detection:** Sobel, Laplacian, and Canny edge detection methods were applied to capture key features.

2.2 Face Detection

Face detection was implemented using OpenCV's Haar Cascade Classifier, a popular and efficient method for face detection based on machine learning and image processing techniques. A pre-trained classifier for frontal faces was utilized, with adjustments to scale factor and minimum neighbor requirements to optimize detection accuracy.

2.4 Skeleton Extraction

The OpenPose model was used to detect and visualize body and hand skeletons. Keypoints extracted were overlaid onto the images to generate skeleton representations, aiding in gesture recognition.

3. Results

3.1 Image Enhancement and Denoising

Image quality improved noticeably after gamma correction and denoising. Mean and median filtering significantly reduced Gaussian noise while preserving edge details, providing a cleaner image for feature extraction.

3.2 Edge Detection

Sobel, Laplacian, and Canny edge detection techniques highlighted different aspects of the images, allowing for thorough feature extraction.

3.3 Face and Gesture Detection

The Haar Cascade successfully detected faces across various scales and orientations. Gesture detection accurately isolated skin areas, which will aid in gesture recognition tasks.

3.4 Skeleton Extraction

The OpenPose-based skeleton extraction effectively identified body and hand poses, providing key points essential for gesture recognition.

4. Conclusion

The presented pipeline for image preprocessing and feature extraction offers a robust foundation for face and gesture recognition. By integrating techniques for image enhancement, noise reduction, face detection, gesture area isolation, and skeleton extraction, the workflow addresses key challenges in preparing high-quality data for machine learning models. Future work can focus on optimizing parameters further and implementing real-time processing.

5. References

- <https://www.kaggle.com/code/serkanpeldek/face-detection-with-opencv/notebook>
- <https://pyimagesearch.com/2014/08/18/skin-detection-step-step-example-using-python-opencv/>
- <https://www.kaggle.com/code/rkuo2000/openpose-pytorch>
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