

UNIK 4690 Computer Vision Project

Pose & Se(kk)g

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Abstract

Introduction

Object Detection is a field within Computer Vision with the purpose of detecting objects of different classes. These classes of objects range from footballs to pedestrians. There are many methods for detecting the latter, but sadly, most of these methods care not to specify the position of a person with high precision. Instead, some methods return rectangular patches where the person resides, while others return only a handful of key-points.

This project aims to combine modern pose detection methods, like [1], with image processing and feature detection methods like Laplacian Blending and Edge Detection in order to not only detect people in images, but to better describe which pixels they inhabit. Once the human's pixels are extracted, they can be used for various things, like 3D reconstruction or transferring them to other images via image blending. The latter of which we have successfully implemented.

The following chapters will discuss what tools we've used and roughly touch on the theory behind them. We'll clarify what we've borrowed from others. Which things we've implemented during our process, both that which works and that which did not. And finally, we point out what we wish to improve and further implement.

Theory

This chapter presents the methods we've used and the theory behind them.

Pose Detection

Pose Detection is a computer vision technology which aims to detect key-points points, often on a person's body, in order to define that object's pose. Pose is in this case the translational information of the person in the image, its position in other words. In image 1 you can observe the algorithm¹ at work. The pose detection algorithm is the basis for our project, and as you can see, it finds key-points on both bodies and faces.

There are many ways to go about pose-detection

¹This algorithm was written by OpenPose
Link to source: <https://github.com/ildoonet/tf-pose-estimation>

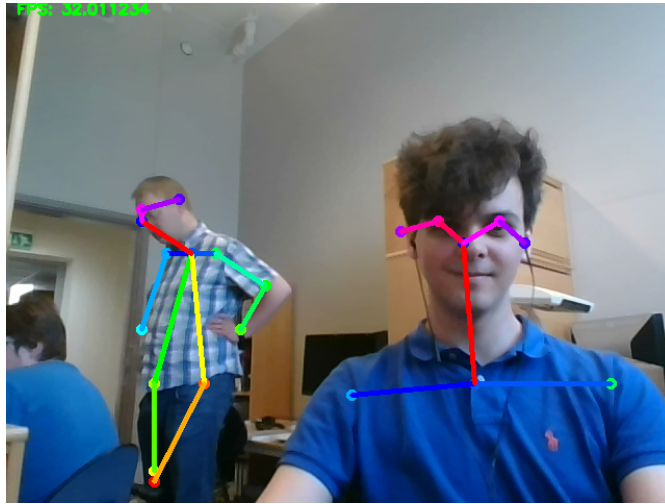


Figure 1: Pose detection algorithm which finds key body-points.

Idea Process, 2D/3D, and Library Independence:

In the beginning we had an idea to try to go into segmentation and explore methods to possibly segment people in 3D. The concept there was to try to find the enveloping hull that closest fit the human being in 3D, using only a single camera and known geometric shapes that approximate the human form.

In order to do that we looked into using Pose-estimators such as OpenPose, and generating Spheres, Cubes, etc. to envelop the person.

On further feedback, we were guided towards doing more image segmentation, and less of the first concept to "scan" a human being into a 3D model. It was even suggested that we would/could try to map the images as we captured them onto a 3D model.

However, in order to do so, we would need to generate a 3D imaging tool which would allow us to both segment, capture, create a mosaic, and use something like ORBSLAM to keep track of the features and keypoints on this expansive surface that covers a human being.

Working on how to implement this, we decided therefore to first get the 2D elements correct. That is; To do segmentation using geometric shapes (Triangles, Circles, Polygons, etc.) and then, if time allowed us, to use these shapes as projected images of the 3D shapes of pyramids, Spheres, etc.

Along the way we would then also need to be able to extract the portions of the image that correlates to the overlap of these shapes and the camera feed. Which essentially turns it into a green-screen (without the green).

Ideally we would have a code that can take any coupled set of Canonical shapes, and a Pose-detection software, map the Shapes to the Pose, and from there be able to extract the information necessary.

The concept should hold for any kind of object, be it a simple or complicated one ideally.

The Process.:

Along the long road towards a usable 2D segmentation we tried a few different approaches.

First out was to use a Canny Edge detection algorithm and try to filter out which edges were closest to any pose-points. After exhausting our paths that way, we determined that it would be better to operate on just a portion of the image. So we rewrote the algorithm to extract regions around points of interest, using element-wise multiplication with Filters.

These Filters were geometric shapes, and we first tried to simply extract the portions of the Canny image that were along the border of our picture. With the goal of using this then with the convexHull-function in openCV.

Our next attempt brought in a bit more interesting twists, as we started playing around with Laplacian Pyramids, and the concept of using the lower tiers of the Laplace Pyramid as our edge-detector. (The concept is not unlike extracting Detail-Spaces using Wavelets)

This turns out to maybe have some uses, but we are still then on a more or less plain segmentation-implementation.

References

- [1] Alexander Toshev, Christian Szegedy.
DeepPose: Human Pose Estimation via Deep Neural Networks. 2014.
http://openaccess.thecvf.com/content_cvpr_2014/papers/Toshev_DeepPose_Human_Pose_2014_CVPR_paper.pdf