

MotionRender: A simple Python implementation of video motion visualization for 3D motion capture data

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Abstract

We describe a Python library project for generating animations of simple 3D motion capture data points. The motion capture method is agnostic with respect to this visualization tool, expecting a time series of time stamped coordinates captured in a 3-dimensional space. One of the advantages of this tool is the simple data format design. This tool was developed originally for a Kinect system implementing skeleton tracking of 15 joint positions, where each data point consists of an accurate time of capture (time stamp), and 3 accurate coordinate (x, y, z) positions of each of the 15 joints at each time step. In addition to the motion capture time series, all that is needed is a joint graph file describing the relationship of the joint edges to one another. This library can be extended to visualize and create animations of motion capture data that can be reformulated using this basic structure of a time series of time stamped 3D coordinates and a joint graph description file. This library should be useful as is, or with easy modification, for many such visualization requirements of similar motion capture data.

Keywords

motion tracking, skeleton tracking, scientific visualization

Introduction

Motion capture devices range from simple consumer grade devices, like the Kinect camera-based depth sensors [1, 2] to more serious motion capture systems [3, 4] with fine spatial and fast temporal resolutions used for generating mocap video rendering.

Creating animated visualization in the base Python 3.X scientific python stack is certainly much simpler than it has been in the past, but still requires many specialized steps to achieve the task of creating an animation that can be saved as a movie for presentation. The standard Matplotlib animation library documentation [5] contains several examples of generated animations, but only 1 of animating a 3D set of data [6]. The basic technique to use an instance of `matplotlib.animation` to generate a sequence of frames from a set of 3D points is clear from the documentation, but can certainly be difficult to generalize for much more complex data than showing the basic example.

3D motion capture data are a very useful type of data that needs to be visualized to analyze its significance. The goal of the `Motionrender` library is to simplify the task of creating animated visualizations of such 3D time series data. The library presented here can be generalized to work with a time series of N 3D points in space to create an animation of subsequent frames of the captured points evolving over time. The only additional general information needed is a matrix of the relationship of the N points in space that should be connected as being related points in the resulting animation, known as a joint graph or joint edge graph by the library.

We used motion capture from a very basic Kinect skeleton tracking experiment, using the NiTE 2.0 library [7, 8] which includes skeleton tracking capture in the development of this visualizaiton library. However the general format, described in methods, can be extended to many kinds of motion capture devices where a timestamp and a set of 3D point positions can be obtained for the captured object in motion. In the example implementation discussed, we have $N = 15$ joint positions (e.g. head, neck, left shoulder, etc.) captured by the motion Kinect depth sensing cameras. An example of a rendered frame of the motion capture animation is shown in Figure 1. Sample videos of the resulting animation are available in the repository for this project, referenced below.

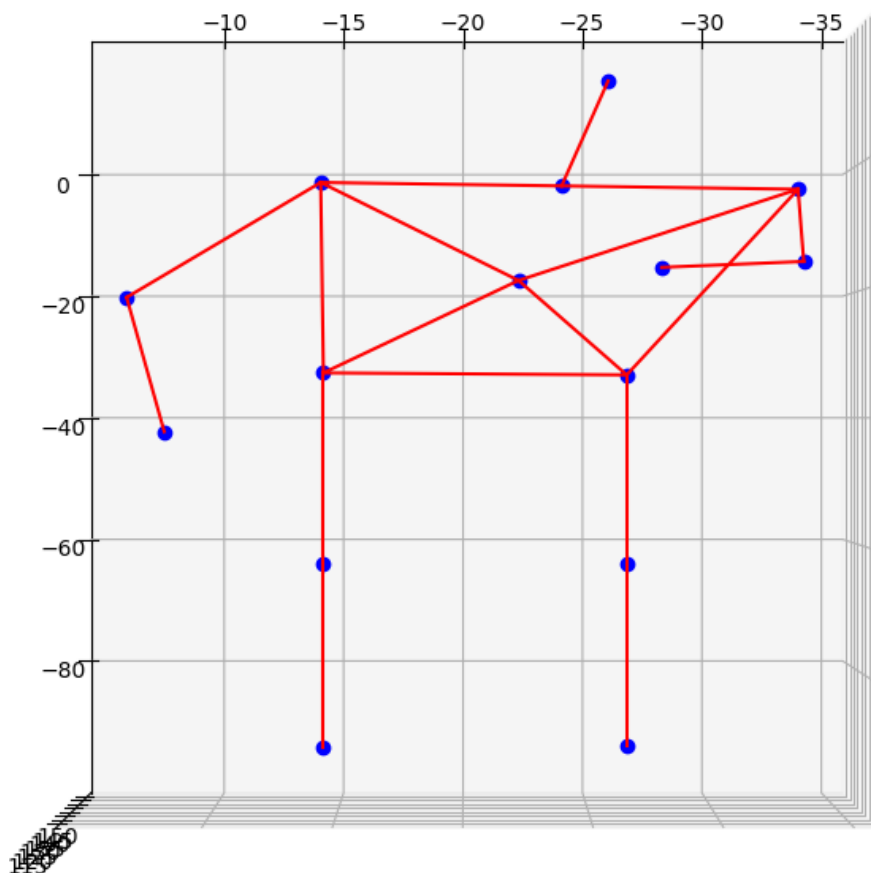


Figure 1. [Example of a rendered frame of 3D motion capture joint positions]

Methods

Implementation

The version of the MotionRender library described here was developed on a Python 3.X scientific tool stack environment. It has dependencies on the pandas and matplotlib python libraries. In addition, for back end rendering, the ffmpeg tool or similar tool is needed.

We have 15 joint positions in the example subject motion captured data sets used in developing the library. Each joint position is captured approximately 30 times per second by the Kinect sensor programmed with the NiTE 2.0 skeleton tracking library software. So while the library was generated specifically with 15 3D motion capture points, the general format of the input data needed to render a 3D animation is

```
timeStamp, jointOX, jointOY, jointOZ,
          joint1X, joint1Y, joint1Z,
          ...,
          jointNX, jointNY, jointNZ
```

as a standard comma separated value (csv) data file, where N can be specified when rendering a motion capture animation of how many joint/position points are in the input. The only other information needed by the library is a joint connection graph of the points if we want to visualize the relationship of the rendered points to one another while being animated. Again using the Kinect 15 joint position data, we specify a joint graph for the software like this:

```
# the relationship of the joint points to assigned numeric position.
# the numeric id corresponds to the expected position in the input file
joint_names = [
    'Head', 'Neck', 'Torso',           # 0 1 2
    'LeftShoulder', 'RightShoulder',  # 3 4
    'LeftElbow', 'RightElbow',        # 5 6
    'LeftHand', 'RightHand',          # 7 8
    'LeftHip', 'RightHip',             # 9 10
    'LeftKnee', 'RightKnee',          # 11 12
    'LeftFoot', 'RightFoot'           # 13 14
]

# joint position graph for visualizing relationship between joints
joint_graph = [
    (0, 1), (1, 3), (1, 4),           # head to neck, neck to shoulders
    (3, 5), (4, 6),                   # shoulders to elbows
    (5, 7), (6, 8),                   # elbows to hands
    (3, 2), (4, 2), (9, 2), (10, 2), # shoulders and hips to torso
    (3, 9), (4, 10), (9, 10),         # shoulders to hips, connect hips
    (9, 11), (10, 12),                # hips to knees
    (11, 13), (12, 14),               # knees to feet
]
```

The user specifies the joint connection graph as a second input to the library for operation. The symbolic names of the joints must match the names in the time stamped motion capture time series of points. For example, the full Kinect connection graph of the 15 captured joints is specified as:

```
head neck
neck leftShoulder
neck rightShoulder
leftShoulder leftElbow
rightShoulder rightElbow
leftElbow leftHand
rightElbow rightHand
leftShoulder torso
rightShoulder torso
leftHip torso
rightHip torso
leftShoulder leftHip
```

```
rightShoulder rightHip  
leftHip rightHip  
leftHip leftKnee  
rightHip rightKnee  
leftKnee leftFoot  
rightKnee rightFoot
```

Operation

MotionRender requires matplotlib and pandas libraries for its operation.

```
$ sudo pip install matplotlib pandas
```

In addition the ffmpeg library is used as a back end frame renderer for creating video files.

```
$ sudo apt install ffmpeg
```

Install the MotionRender library from standard PyPi python library.

```
$ pip install motionrender
```

The library expects a standard csv file of values, where the first column is a time stamp. Subsequent columns are the data for motion capture point 0 (x, y, z), motion capture point 1, etc. So given 15 motion capture points, this library expects an array with 46 columns, where the first column is a time stamp, and the subsequent columns are the 3D positions for all points captured. This could be loaded from a csv file, or generated from some other source.

The second parameter corresponds to the joint graph shown above. This is expected to be a space separated file of edge relations defined between the joint position names. The joint position names for the time series capture data should match the joint position names in the joint graph. The library currently contains two main API functions that may be used. The `render_frame()` member function of the main `MotionRender` class renders single frames from the motion capture data. While the `render_animation()` function renders video files. Actually the default behavior of both of these methods is to return `figure` and `animation` object instances respectively. However, the user can supply a file name to these API calls and the figure/video will then be saved in the specified file. Following is a quick example of using the library interactively to read in motion capture data, and render it as figures and videos.

```
$ pip install motionrender  
Collecting motionrender  
  Downloading motionrender-1.0.0-py3-none-any.whl (18 kB)  
Installing collected packages: motionrender  
Successfully installed motionrender-1.0.0  
  
$ python  
Python 3.9.7 (default, Sep 16 2021, 13:09:58)  
[GCC 7.5.0] :: Anaconda, Inc. on linux  
Type "help", "copyright", "credits" or "license" for more information.  
>>> from motionrender import MotionRender  
>>> mr = MotionRender("data/standing-subject.csv", "data/standing-joint-graph.csv")  
  
>>> fig = mr.render_frame(1636576712852000, figure_name="standing-subject.png")  
  
>>> anim = mr.render_animation(movie_name="standing-subject.mov")  
processing frame: 0  
processing frame: 500
```

Additional files are included in the main directory of the example project, that includes the data file use cases shown here. The test files contain additional examples of configuring rendering options of the library. You can invoke them in the example project as:

```
$ python test-plot.py  
$ python test-render.py
```

In addition there is also a jupyter notebook present in the example project named `test.ipynb` that shows further how the library can be used interactively.

Use Cases

As a use case, we use the `MotionRender` library and generate animations of a set of Kinect motion capture data. This data was used in the analysis of [9] research paper. This data and the software workflow to render participants in the experiments can be found in the source code repository listed below. Two samples of the kinect motion capture data, from a sitting and standing participant respectively, can be found in the example project. These files give further examples of the data format expected by the input files for the library.

Limitations

This library relies on current version of Python and matplotlib animation library. Data is expected in the needed format, which may not always be simple to massage into depending on the motion capture device output. A rendering background tool, such as `ffmpeg` must also be installed and accessible to `matplotlib`.

Conclusions

The `MotionRender` library documented here provides a small tool that may be of use to those with motion capture data, or similar data sets that consist of a time series of points captured at intervals. The tool uses a simple data format, needing a flat file of captured points and another file defining the point graph relationships for rendering. The python `matplotlib` animation facilities are a bit hard to use. It is hoped this library would fill a gap and allow for easier rendering of motion capture data.

Data availability

The motion tracking dataset used to develop the software described here can be found in the archived example user code data repository. The csv files for this dataset can be directly downloaded from the GitHub repository: <https://github.com/DerekHarter/motionrender-example>

Software availability

- Source code of user example, use cases and example data of the `MotionRender` library is available from: <https://github.com/DerekHarter/motionrender-example>
- Archived source code of the PyPi published package at the time of publication: <https://test.pypi.org/project/motionrender/1.0.0/>
- The development repository for this library, if you are interested in the code internals, can be found here: <https://github.com/DerekHarter/motionrender>
- License: GPL-3

Competing interests

No competing interests were disclosed.

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