**Motion capture data processing and analysis**

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# List of Abbreviations

3D Three-dimensional

MoCap Motion Capture

fps frames per second

SL Sign language

HamNoSys Hamburg Notation System

RP Rest Pose

# Introduction

Motion capturing is a modern, fast developing data acquiring method capable to record movement in 3D. The detailed data that is retrieved from such recording is very useful not only for movie and game industry, but also in other fields as military, medicine and for validation and control of computer vision and robotics. The benefits of this method attract scientist to utilize it for linguistic analysis. Sign language is the primary alternative to a spoken language. Unfortunately, there are people for whom this is the only alternative that gives them the ability to communicate and share thoughts directly. Sign language uses manual movements and body language to communicate thoughts with others. The basic component of a sign language includes hand gestures, movements, orientation of fingers and hands, hand shapes and facial expressions to communicate certain feelings. Every region in the world has a unique spoken language and similarly, every region has a unique sign language. Thus, sign language varies from culture to culture and from region to region. People with speech and/or hearing impairment find it difficult to communicate with other individuals via sign language due to the inability of most of the people to understand sign language.

The purpose of this paper is to show the methods of studying and understanding the properties of signs from motion point of view and developing a tool for processing sign language data base. It is focused on hand movement and gestures analysis. It is cultural and regional independent, because it uses kinematic and statistical methods for processing the data.

Such database is set of dictionary (lexical items separated by default stance //not sure if it should be here or later in the main part) files with motion capture data. The tool will perform raw segmentation (in the first step) and fine segmentation (using acquired parameters from data) of dictionary items to extract the meaningful information for each sign which will be used for further analysis. It will show the challenges to determine the exact beginning and ending of the sign. The significant problem with the nature of the data – the containing of noise and methods for isolating that noise. Each sign will be processed for extracting its properties. Such as: if the sign is one handed or two, which is the dominant hand, hand location and orientation, finger orientation.

Later with the help of computer learning methods such as SVM the extracted information will be used to cluster signs for further processing.

# Theory. Analysis. Aim.

## Motion Capture

### History of motion capture

In general, the term motion capture (MoCap) is understood as the process of recording the movement of object, people or animals. It is not specifically related to any device or approach. Today’s motion capture systems are product of many years of tinkering and innovation.

At the beginning MoCap analysis originate as gait analysis and animal locomotion. A pioneer in this field is the photographer Eadwerd Muybridge. In his work he used multiple cameras, triggered by strings to take pictures of moving bodies and animals. He manages to capture what human eye could not distinguish as a separate movement. His work “The horse in motion” was the first work recognized as a motion capture analysis. With a series of photographs of a galloping horse (see figure 1) he proves that horses do have all four hooves of the ground during their running stride [[3](file:///C:\Users\rusev\AppData\Roaming\Microsoft\Word\Muybridge's#_E._Muybridge,_)].

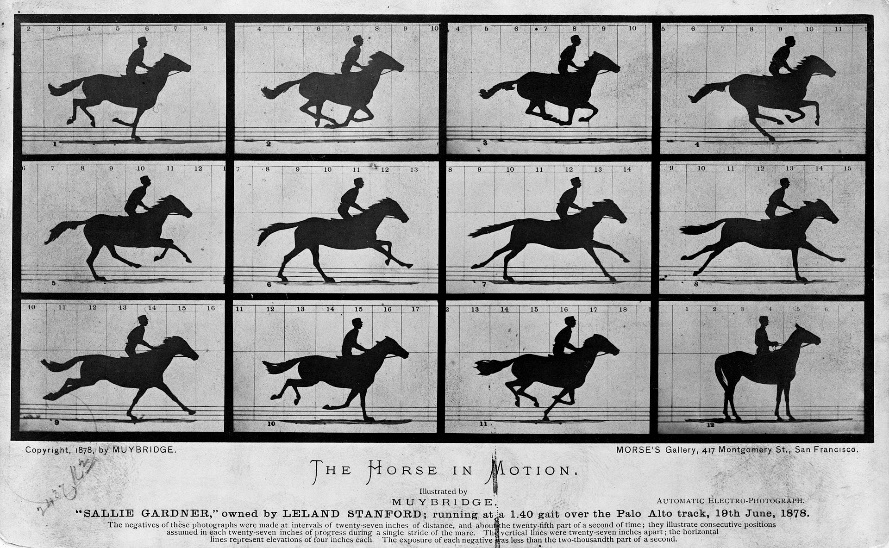


Figure 1: "The horse of motion" by Eadwerd Muybridge

Since then the technology has developed to the form of devices designed for direct 3D recording. The biggest success and from where it gains its popularity is the usage in the movie industry. Jar Jar Binks is the first fully digital created character with the help of MoCap. It was performed by Ahmed Best in Georges Lucas’ Star Wars Saga in 1990.

Motion capturing techniques started as biomechanics research tool, gained its popularity with its usage in entertainment industry and developed and expanded into education, training, sports, and robotics.

There are different approaches for motion tracking. The most advanced technologies are capable of tracking motion with high precision at very high sample rates. Optical passive system is considered as one of the most accurate and flexible technique and thus very common in the industry. On the other end of the scale are low-cost sensor technologies that most people use daily in their mobile phones, laptops, game controllers and so on. Every system has its advantages and disadvantages depending on its purpose.

The technologies presented in this section include acoustic, mechanical, magnetic, inertial and optical tracking.

**Acoustic tracking** systems calculate position upon the wavelength of an acoustic signal and the speed of sound. This approach is vulnerable due to the fact that sound speed varies with air pressure and temperature.

**Mechanical tracking** systems are usually based on some mechanical construction which measures angles or lengths between mechanical parts. These systems may be worn on the body, for example implementing sensor in exoskeleton or a glove. Mechanical tracking has been popular in music-related work, particularly for the purpose of developing new musical interfaces.

**Magnetic tracking** systems use the magnetic field around a sensor. Passive magnetometers can measure the direction and strength of the surrounding magnetic field. More advanced magnetic systems use an active electromagnetic source and a sensor with multiple coils. The position and orientation of each sensor can be calculated as a function of the strength of the induced signal in each sensor coil. Magnetic trackers are able to operate at high sampling rates (more than 200 Hz) with high theoretical accuracy. However, the systems are sensitive to disturbances from ferromagnetic objects in the tracking area.

**Inertial tracking** systems include those based on accelerometers and gyroscopes. These sensors are based on the physical principle of inertia. Inertial tracking systems have certain strong advantages. Firstly, they are completely self-contained, meaning that they do not rely on external sources such as acoustic ultrasound sensors or cameras which require line-of-sight. Secondly, the sensors rely on physical laws that are not affected by external factors such as ferromagnetic objects or light conditions. Thirdly, the sensors are very small and lightweight, meaning that they are very useful in portable devices; and finally, the systems have low latencies and can be sampled at very high sampling rates.

Various types of video camera are used in **optical motion tracking**. In principle, any digital video camera can be used — in fact, one of the most affordable sensors for conducting motion tracking is a simple web camera. Cameras used in optical motion tracking are either (1) regular video cameras, (2) infrared (IR) video cameras, or (3) depth cameras.

Ordinary video cameras sense light in the visible part of the electromagnetic spectrum. Each pixel in the camera image contains a value corresponding to the amount of light sensed in that particular part of the image.

Infrared cameras sense light in the infrared part of the electromagnetic spectrum. Some infrared cameras can capture heat radiation, e.g., from humans, but the most common use of infrared cameras in tracking technologies is in a slightly higher frequency range. This is achieved by using some active infrared light source, and either capturing the light from this source directly or as reflections on the tracked objects. Typical implementations consist of a group of infrared light-emitting diodes (LEDs) positioned near the infrared camera and capturing the reflection of this light as it is reflected from small spherical markers.

Depth cameras provide a layer of depth information in addition to the regular two-dimensional image. These cameras use some technology in addition to the regular video camera. When not provided by the camera itself depth information can be gained through the use of stereo cameras. This involves two cameras mounted next to each other, providing two similar images. Depth information is found as a correlation function of sideways shifting of the images.

After obtaining the video data various processing is applied to the video stream. The video processing that is performed in optical tracking systems is primarily dependent on two factors: (1) whether or not the tracking is based on markers and (2) the camera configuration. But in any case the first processing step is to remove unwanted information from the video, i.e. separate the foreground from the background.

### Optical-based motion capture system VICON

Motion capture technology was used to generate data files for both continuous and isolated utterances. The data used for this project was recorded at the University of West Bohemia with MX series device from VICON [[6](file:///C:\Users\rusev\AppData\Roaming\Microsoft\Word\Go_Further_with#_)]. This is an overview of the basic principle of the technology on which the system is based. This system is based on optical- passive technology and was chosen for the sign language project because it suits the best for the purposes of the project.

The technology provides accurate data at fast sampling rates, and the same system can be used to capture simultaneously the motion of a wide range of structures, including objects, animals, human bodies, fingers and faces. By using passive (reflective) markers, all processing is done externally, and the captured subject does not need to wear electrical equipment or wires. Which can be considered as advantage in capturing finger motion, because the presence of wires can impede the naturalness of movements. The system is comprised of eight specialized infra-red cameras along with computers and software for image analysis and processing. The cameras detect small markers placed on strategic locations on the captured subjects. Markers are treated as points without volume, and as such only their positon (not orientation) can be tracked. On figure 2 can be seen the general marker set and marker’s placement for human body. For this project passive markers are used, they are coated with retroreflective material, and this requires the cameras to emit the light, which is reflected back and detected. The external source of light is part of the camera’s body.



Figure 2: General marker set for human body

The detected information is then processed in dedicated hardware using software provided by the manufacturer to triangulate the 3D locations of the markers. The exact process and algorithms are know-how of devices’ manufacturer, but it is based on stereoscopic vision.

After reconstructing the 3D points into a point cloud, the system needs to determine which point is which and label each point with a marker id. This process is commonly referred to as marker labelling. The marker data is then used to estimate the kinematic motion of a model of a human skeleton. The estimation in form of bone lengths and joint angles can be used for further processing or animation. This process is called solving. The mentioned software is able to provide semi-automatic calibration of human subject for general movement.

Although this technology comes with highly accurate results and it is very flexible there are some disadvantages which may result in poor data quality or extensive costs in manual post-processing. As every vision-based technology it needs clear line of sight and occlusions may cause serious challenges. Markers placed on fingers are especially problematic and often suffer from self-occlusions when the fingers are bent or the hands are facing towards the body or with palm-up [[5](#_N._Wheatland,_Y.)]. Further challenges arise in situations when several markers come in close contact (such as clapping hands).

To summarize, optical motion capture provides highly accurate data, but may require a large amount of manual post-processing.

## Sign language analysis

Sign language is used by millions of people around the world. It is used to facilitate communications with people with speech or hearing impairments. There are two common misconceptions: 1) Sign language is one and same everywhere; and 2) sign language is dependent on spoken language [[8](#_Perlmutter,_David_M.,)]. But the truth is that there are different sing languages as there are different spoken languages. Linguists has studied and proved that sign languages exhibit the fundamental properties that exist in all languages and there are similarities between both forms, but there are some basic differences. The linguistic mechanism in both is different and therefore it causes difficulties for people with such impairments, especially for those who are born this way, to use even the written form of a spoken language [[9](#_Sandler,_Wendy;_&)]. In spoken language units are organized sequentially (it is not possible to say two different words at the same time), but in sign language the meaning of one unit may be carried by the shape of the hands and their position or/and by the position and movement of head and mimics. These two components can be carried simultaneously.

All these circumstances not only put a communication barrier between people using sign language and majority of hearing community, but also restrict them from most sources of information. Another difference between both is that sign languages does not have its own, unified notation system. This led scientist to develop writing system to represent signs. The pioneer in sign language analysis was W. Stoke. As it is explained in M. Kato paper [[3](file:///C:\Users\rusev\AppData\Roaming\Microsoft\Word\A#_Mihoko_Kato,_)] Stoke proves that each sign in American sign language has tree elements that distinguish it from all other signs:

1. Hand Configuration (the distinctive configuration of the hand or hands making a sign),
2. Place of Articulation (the place where a sign is made),
3. Movement (the action of the hand or hands).

Stokoe decided to call the active hand the “designator” or “dez”; the place, the “tabula” or “tab”; and the action, the “signation” or “sig.” A sign is produced by a combination of these three aspects.

Nowadays most of the notion systems for sign languages are based on his study and notion system.

### Hamburg Notation System

The Hamburg Notation System (HamNoSys) [[1](file:///C:\Users\rusev\AppData\Roaming\Microsoft\Word\Hamnosys-representing#_T._Hanke,_)] is a work of scientists from University of Hamburg. It is an alphabetic system that decompose signs to phonetic level and describes their sub lexical parameters such as location, configuration and movement. It is based on Stokoe’s notation system [[2](#_Stokoe,_William_C.)]. It is designed to be usable in variety of context with the following goals in mind:

* International use - HamNoSys transcriptions to be possible for virtually all sign languages in the world.
* Iconicity – because of the large number of parameters variations, newly created glyphs should be designed to be easy to memorize or even deduct the meaning
* Economy – notation of signs should make use of principles as symmetry conditions, thus resulting in shorter notions.
* Integration with standard computer tools – It should be usable for computer –supported transcription, standard text processing and database applications.
* Formal syntax – The notation language should have well-defined syntax.
* Extensibility – As SLs are developing and differ from each other, HamNoSys should allow both for a general evolution and specializations. [[1](file:///C:\Users\rusev\AppData\Roaming\Microsoft\Word\Hamnosys-representing#_T._Hanke,_)]

Because of all described goals of HamNoSys it seems that it is good choice for scientific point of view and preferred from scientific community working in the domain of language analysis and synthesis, although it is not very accepted from deaf community.

This notation system was chosen for sign language project because it suits the best for its purposes. I used it as guidelines in my work for describing the properties of signs from motion point of view.

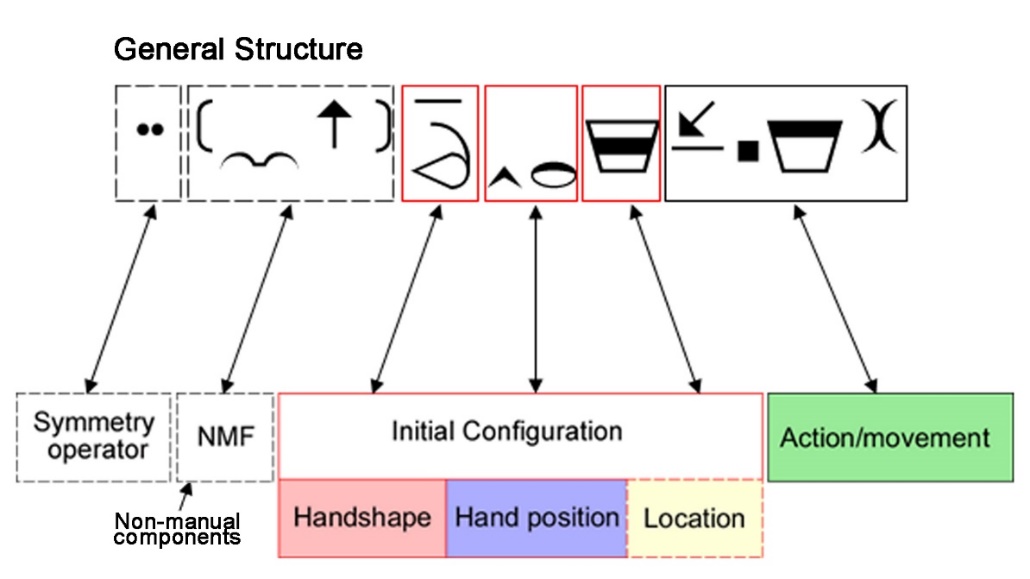


Figure 3: General structure of HamNoSys notation

In general, a sign notation consists of description of non-manual features, handshape, hand orientation and location, plus the actions changing this posture. If the sign is two handed at the beginning of notation is added an operator to show how the description of dominant hand is copied to the non-dominant hand. The signs are realized in signing space and terms expressing directions are determined from the signer’s perspective. Figure 3 shows example of sign notation and its general structure with its mandatory (the ones with solid border) and optional components (boxes with dashed border).

#### Handshapes

Handshape is expressed by symbols for basic forms – **Fist, Flat hand, Separated fingers, Thumb combination** and bending (see Figure 4)

The configuration of the thumb alters the structural arrangement of the entire hand and thereby define a new group of handshapes, where shapes are derived from other basic shapes.

In addition to the general description of the sign, information for different fingers and finger parts (in respect to the fingers involved), may be included.

Figure 4: Handshapes

#### Hand orientation

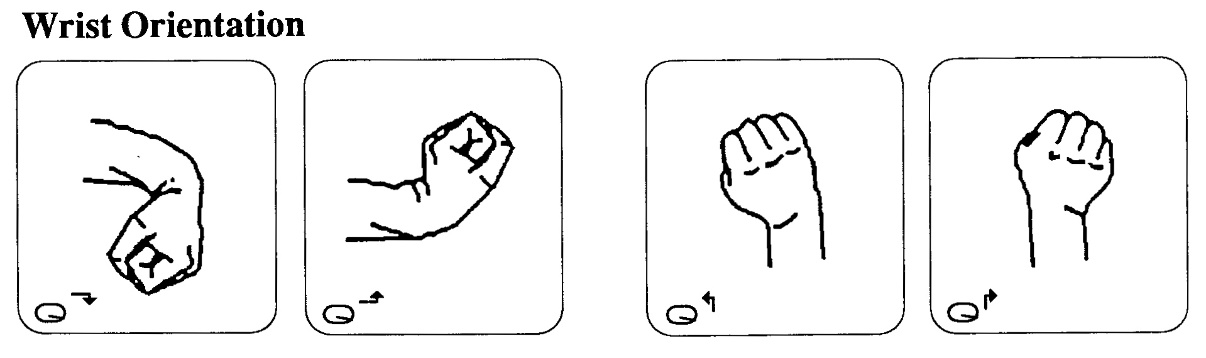
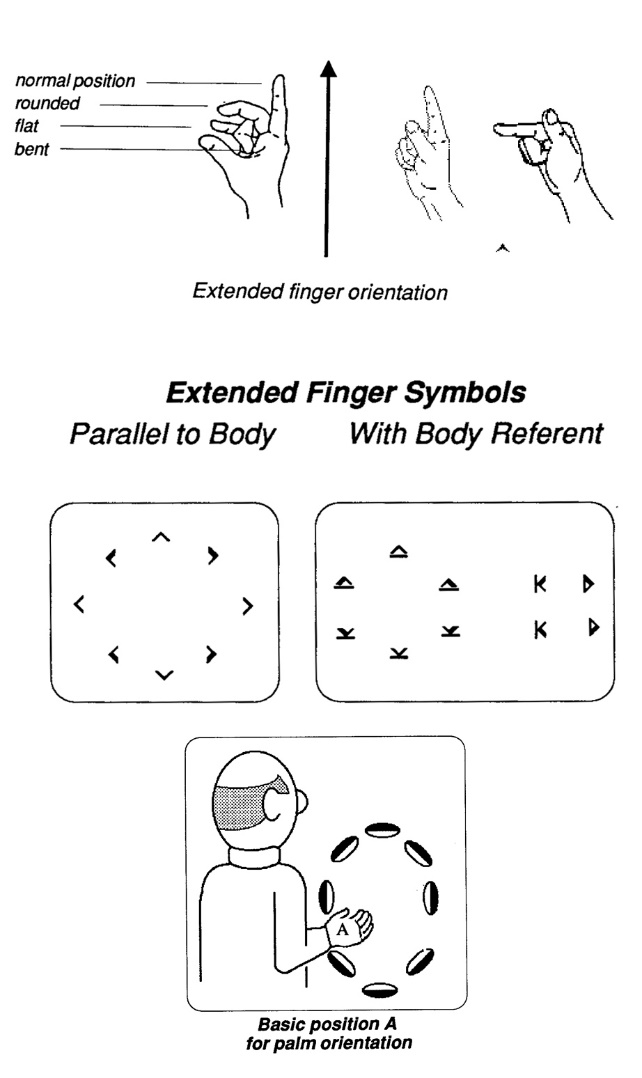
The orientation of hand is described by **Wrist orientation, Extended finger and Palm orientation**. Where wrist orientation is characterized by bending of the wrist toward the pulse or back of the arm, toward the thumb or little finger (see Figure 5). It is omitted when the bending is natural consequence of palm orientation and movement.

Figure 5: Wrist bending

The vector originating at wrist, running along the back of the hand and pointing to the direction pointed by the fully extended finger, shows the orientation of the Extended finger. This orientation may be difficult to be defined when all fingers are bend in some way. As it can be seen on Figure 6, the two hand shapes on the right are noted with same symbol. Fingers point in different direction, but the base of the finger (knuckle) points in the same. If finger orientation is not parallel to the body plane, then the notation must include reference to the body (Figure 6).

Palm orientation is always noted after the Wrist and Extended finger orientations are. To define it, first Basic Position of the palm must be determined, then the orientation is determined by the orientation of the palm around the shaft of the hand (Figure 6, bottom). There two Basic Positions. When Extended Finger orientation is away from the body is Basic Position A, otherwise is Basic Position B.

Figure 6: Extended finger orientation and Palm orientation

#### Hand location

Hand location is noted only if it is very specific on the body or in the signing space. If not noted it is understood that sign takes place in the natural space, in front of the upper part of the body. There are different sets of location signs. One is expressing locations on body (Figure 7), another one describes relative position to the body.

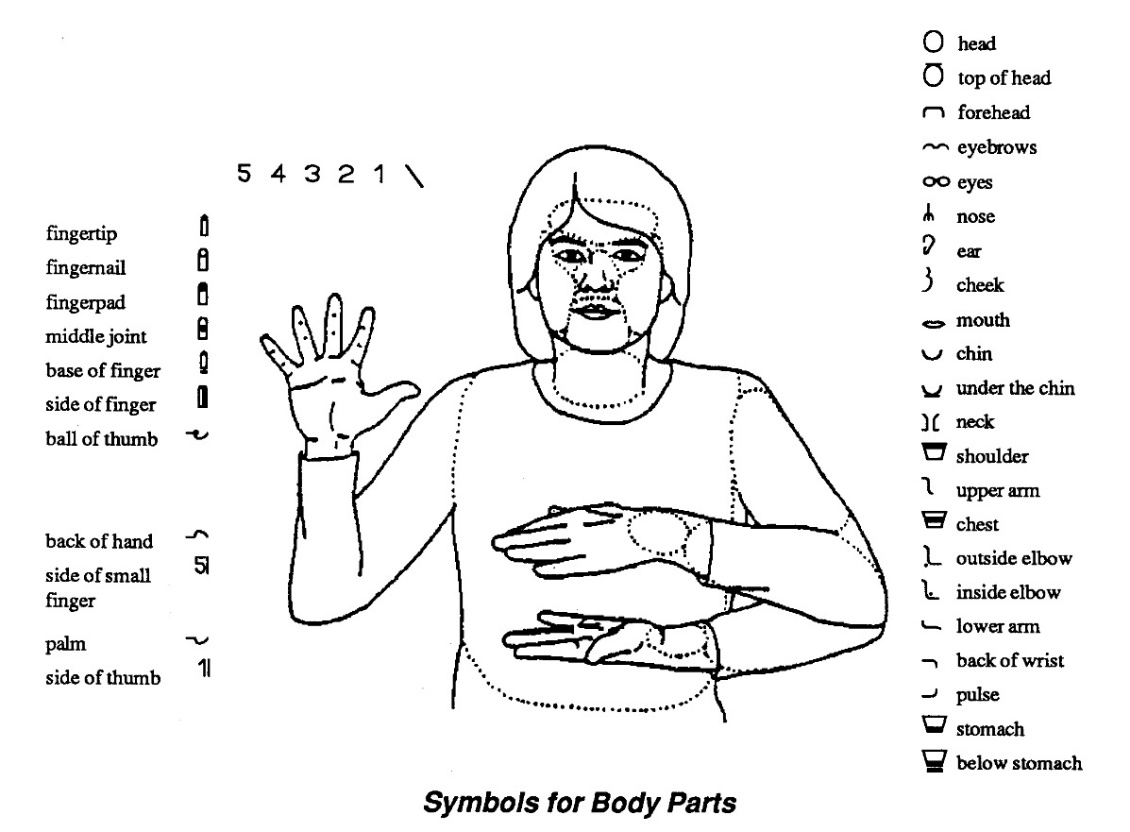


Figure 7: Body parts

Figure 8 shows the division of the signing space into six horizontal zones, each zone is noted with different symbol. In two-handed signs location information concerning hands and arms, refers to locations on non-dominant hand and characterize the relation between the two hands. Again all determinations of left and right are made from signer’s point of view.

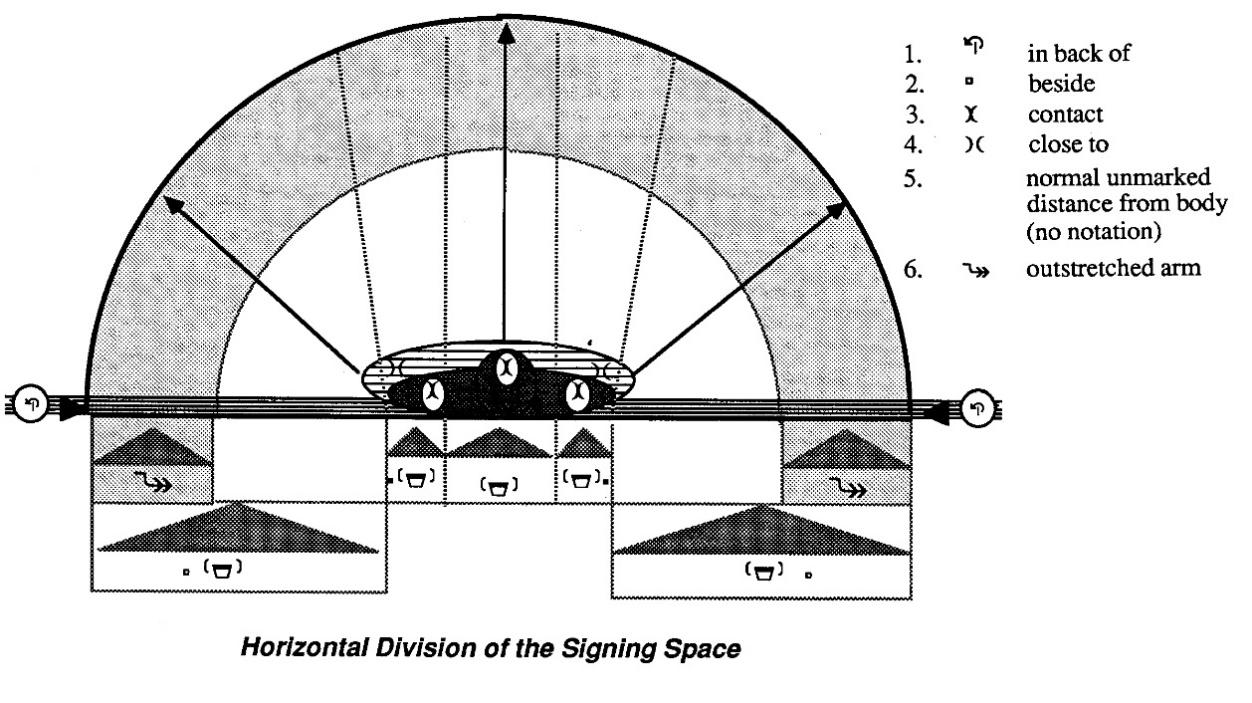


Figure 8: Six zones of signing space and their symbols

#### Movement

There are few aspects of the movement that have to be considered when describing movement in signs: **Movement type, Manner of movement, Repetitions, Order of movement symbols**. The distinguished types of movement are: straight, curved, wavy, zigzag, circular and spiral.

The manner of movement aims do describe size, speed and intensity of the movement. There are three degrees of size: large, normal and small and only large and small size are explicitly notated. The speed and intensity are described as: fast movement, slow movement, tense, hold or rest (usually on onset or offset of the movement), abrupt halt at the end of the movement.

Two types of repetitions are distinguished. Ones which lead back to the beginning of the movement and those whose initial position continuously change. The number of repetitions is indicated by the number of repeating symbols.

The order of notation of these aspects is strongly defined as follows:

1. Circling movement or direct movement
2. Type of movement:
3. Manner of movement
4. Repetition

#### Two handed signs

Two handed signs are divided into symmetrical and nonsymmetrical signs. In the symmetrical ones both hands show the same handshape and therefore only the dominant hand is notated. But the coordination of both hands should be noted either as mirrored or parallel. Also should be noted if the movement in both hands occurs simultaneously or alternately. In case of nonsymmetrical signs, the movement occurs only with the dominant hand and the handshapes are often different, therefore notation for both hands is needed. Always the notation of the dominant hand should precede the non-dominant.

#### Conclusion

Sign language and spoken language developed separately and this led to different linguistic structures, restricting people with haring and speaking impairments from vast information and free communication. Linguist have been working on developing good sign representation as it is important from many aspects. Although it is not very accepted by deaf community, because of its complexity, the Hamburg Notation System seems to be a good choice for representing signs in computer, as it was developed with the idea to be available for international use and easy for computer implementation.

## Optical-based motion capture for sign language synthesis

There have been several research projects for developing assistive technology for Deaf population. They are based on different techniques, such as key frame techniques and procedural synthesis. As it is explained in [[13](#_L._Naert,_C.)] they allow fine control over the movement of the signing avatar, but are not very well accepted by people, because of lack of human-like movements. Scientists as McDonald et al [[4](#_J._McDonald,_R.)] work on resolving this problem by analyzing noise in motion capture data and adding the human-specific noise in the key-frame driven motion. Other approach is data-driven synthesis. When used with MoCap data from real signer the realism in the motion is preserved and therefore improves the animation.

On the other hand, this approach needs a rich MoCap dataset for sign language, which faces other challenges. As mentioned before SL is a complex composition of simultaneous movements of different parts of human body. The difficulty is to record and synchronize all these components, because every utterance can be done with time and special variance even if performed by one signer. At university of West Bohemia, where I had the opportunity to work, deal with this problem by using the state-of-the-art device for motion capturing [[12](#_P._Jedli_cka,_Z.)]. The system they use is combination of eight cameras and set of 109 retro-reflexive markers with different sizes for different body parts. As this is a new approach, the data set is not that rich yet.

Data sets for SL may differ in their content. Some contain isolated utterances other whole sentences or phrases with continuous utterances. For the purpose of my work I have used the dataset with isolated utterance.

## Segmentation of SL

The benefits of data-driven methods for SL synthesis were discussed in the previous section, but in order to have good and valid results the data must be reliable. Before proceeding to any kind of analysis the data need to be structured in forms that suits the final purposes.

Segmentation is the process of breaking a continuous sequence of movement data into smaller and meaningful components. The process consists of determining the exact beginning and ending moment of this meaningful component. Identifying of segments is challenging task due to multidimensional nature of SL.

There is a study over French sign language [[13](#_L._Naert,_C.)] which is based on manually segmented and annotated data. Segmentation is performed by expert annotators (deaf signers). As it is explained in the paper manual segmentation and annotation is laborious and time-consuming process. Identifying the exact start and end frame made by human is subject to variability, even if made by experts. The focus of this study is *transition* movement in the sign. *Transition* is the motion between end of one sign and beginning of next sign. This movement have no linguistic meaning, but it is important for achieving human-like animation.

There are several studies researching the issue of segmentation, but as far as I am aware, there still have not been developed a fully automatic segmentation of continuous utterance with good precision. The approach for identifying segment boundaries by detecting changes in kinematic features for general motion and in the case of isolated utterance seems promising.

### Method for segmentation

In my work the segmentation is designed to be made over sets of isolated utterance and is done in two steps. In these data sets signs are separated with one exact pose – rest-pose (RP). The identification of segments is based on kinematic analysis of hands movement.

At first step start and end tag define these frames where hands leave and enter, respectively, rest-pose. Knowing this frames helps me to analyze hands behavior during rest pose and thus refine threshold variables for finer segmentation on next step.

At second step the frames that I am interested in are those where the meaningful part of sign begins and ends. Meaningful part is the one that have linguistic meaning and is described by HamNoSys notation. This motion subsequence will be further analyzed.

The main idea behind kinematic analysis is that significant changes in trajectory of a hand can be indicated by changes in velocity and acceleration. Indeed, human body is not a simple mechanism and hand movement (or in our case change in the trajectory of the marker) is related to the kinesthetic muscle response, resulting in changes in velocity and acceleration. The correlation between this points of interest and changes in kinematic features was studied at the university of California for the for a project for mapping music to a dancer’s motion [[16](#_F._Bevilacqua,_J.)] and also in a research of a French sign language [[13](#_L._Naert,_C.)].

## Conclusion

Sing language is important for certain part of our community. It is complex and by its nature very different of the spoken language that most of us are familiar with. There have been different researches and projects for developing signing avatar with increasing quality over time. New approaches include data-driven methods, using devices for recording 3D movements in order to save human-like motion. But the issue of automatic processing and computer understanding are still challenging tasks. SL can be reviewed as continuous stream of motion, with no consideration of linguistic meaning of the signs. Therefore, it can be decomposed to phonetic level and described using HamNoSys notation. In order to have valid result proper segmentation of the data is needed, only the ‘meaningful’ motion has to be extracted. Defining start and end moment of the sign is variable task, even performed by experts. But these moments might be detected by analyzing hand kinematics of the motion as position, velocity and acceleration

# System Architecture

After analyzing the theory behind motion capture technology, sign language and its properties and different approaches for sign language synthesis is possible to proceed with actual describing of the system’s architecture. The system consists of two general modules- segmentation module and module for a sign analysis. The tool will perform two step segmentation over input data. The input data is long stream of 3D recorded signs performed by professional signer. After that the motion data will be analyzed to extract motion features of each sign.

## Python

Python is a high-level, interpreted and general-purpose dynamic programming language that focuses on code readability. The syntax in Python helps the programmers to do coding in fewer steps as compared to other programming languages. It is very flexible language supporting multiple programming paradigms, including object-oriented, imperative, functional and procedural. It has a large and comprehensive standard library that has automatic memory management and dynamic features. The programmers of big companies use Python as it has created a mark for itself in the software development with characteristic features like-

* Interactive
* Interpreted
* Modular
* Dynamic
* Object-oriented
* Portable
* High level
* Extensible in C++ & C

One of its biggest advantages is its Extensive Support Libraries that include wide range of areas such as string operations, Internet, web service tools, operating system interfaces, protocols, data-analysis and more. These libraries make Python more useful for specific purposes. Additionally, Python has become a go-to language for data analysis. With data-focused libraries like NumPy, and matplotlib, making it powerful tool for processing, manipulating, and visualizing data [[14](#_Python_documentation_-)].

### **NumPy**

The most fundamental package, around which the scientific computation stack is built, is NumPy (stands for Numerical Python). It provides an abundance of useful features for operations on n-arrays and matrices in Python. The library provides vectorization of mathematical operations on the NumPy array type, which ameliorates performance and accordingly speeds up the execution.

### SciPy

SciPy is a collection of mathematical algorithms and convenience functions built on the NumPy extension of Python. Its different submodules correspond to different applications, such as interpolation, integration, optimization, image processing, statistics, special functions, etc.

### Matplotlib

Another SciPy Stack core package and another Python Library that is tailored for the generation of simple and powerful visualizations with ease is Matplotlib. It is a top-notch piece of software which is making Python (with some help of NumPy, SciPy) a cognizant competitor to such scientific tools as MatLab or Mathematica.

However, the library is pretty low-level, but with a bit of effort, you can make just about any visualizations:

* Line plots;
* Scatter plots;
* Bar charts and Histograms;
* Pie charts;
* Stem plots;
* Contour plots;
* Quiver plots;
* Spectrograms.

There are also facilities for creating labels, grids, legends, and many other formatting entities with Matplotlib. Basically, everything is customizable.

### C3D

This is a small library for reading and writing C3D binary files. C3D files are a standard format for recording 3-dimensional time sequence data, especially data recorded by a 3D motion tracking apparatus.

## Sublime Text code editor

Sublime Text 3 is a lightweight, cross-platform code editor known for its speed, ease of use, and strong community support. It’s an incredible editor right out of the box, but the real power comes from the ability to enhance its functionality using Package Control and creating custom settings. There are many plug-ins accessible that make Python development extremely smooth and pleasant.

**Anaconda** is an extremely powerful Python package for Sublime. It offers:

* Python code auto completion
* Python lining (highlights both syntax errors and PEP8 violations)
* Python documentation

## Autodesk MotionBuilder

should I include it?

## C3D file format

C3D is a biomechanics and motion capture file format. It stores raw 3D coordinates and analog sample data, together with information that describes the stored data. The C3D format treats information as if it belongs to one of two classes: **Physical Measurements, Parameter Information**.

Physical Measurements - The C3D specification expects physical measurements to be one of two types, either positional information (3D coordinates) or numeric data (analog information).

Parameter Information - contain information about the data such as measurement units and data point labels, database information such as the subjects name, diagnosis and other items that may be specific.[[15](file:///C:\Users\rusev\AppData\Roaming\Microsoft\Word\The_C3D_file#_)]

For this project C3D file format is used for the input data. It contains the raw output data of the MoCap system.

## Activity flow

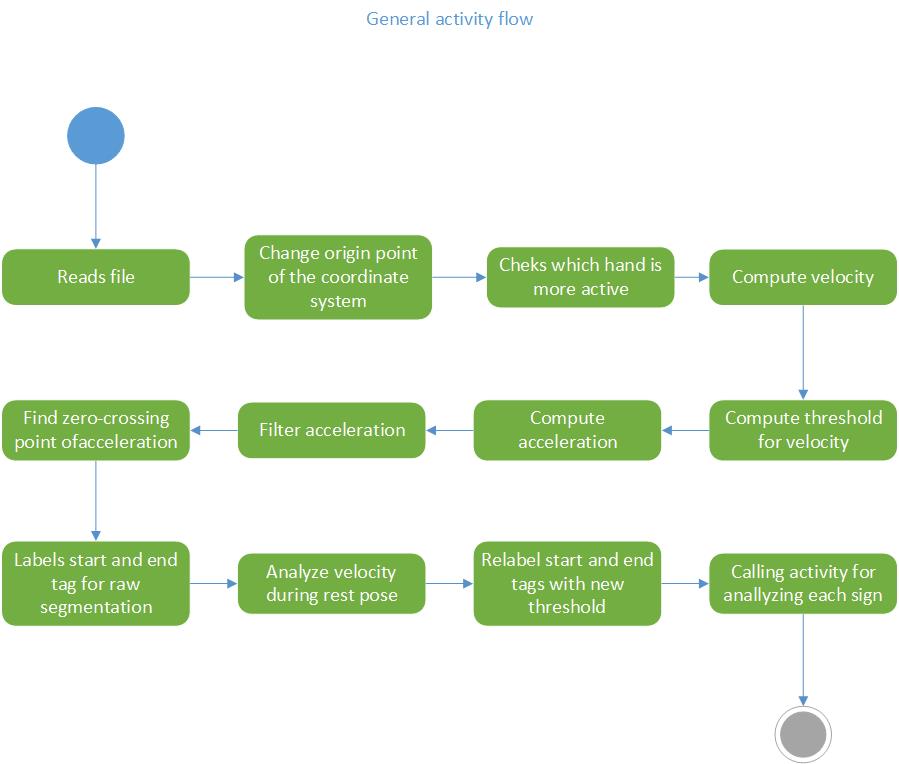
The dynamic behavior of the system is shown in the following UML activity diagrams.

Figure 9: General activity diagram

The general activity flow of the system is shown on Figure 9. First the input file is read and the information about markers positions is stored in multidimensional array (in form of: for each frame, for each marker, x, y, z, coordinates). Other extracted information is the frame rate and the set of labels of used markers (in a list structure). The last element of the list of markers is additionally added marker for the origin of the coordinate system.

Next activity is to check which hand is more active. The further analysis is performed over the movement of the dominant hand.

Next step is to recalculate 3D co-ordinates for each frame so that the center of the co-ordinate system to be in the middle of the body. This is done in order to make further calculations independent of the signer’s physicals characteristics or his position in the scene.

After that the hands velocity and acceleration are computed. Acceleration computation is made over the normalized velocity.

A crucial step in the processing of the data is the use of low pass filtering to eliminate jitter in the acceleration. It is important because of the next step – finding the zero-crossing points. If the function is too noisy there might be found false points. On the other hand, filtering cause loss of information and thus missing important points. As explained earlier the sudden changes in velocity indicates changes in trajectory. Therefore, this points are matter of interest, they can indicate a start or end of a sign.

Next step is called labeling. It uses a threshold function to filter the set of zero-crossing points. The result at the end of this step is list of start and end frames for the raw segmentation.

Knowing this start and end tags, gives the knowledge to analyze hand velocity during rest-pose and this way to choose better value for the threshold function.

Next step is just redoing the labeling with the new threshold value.

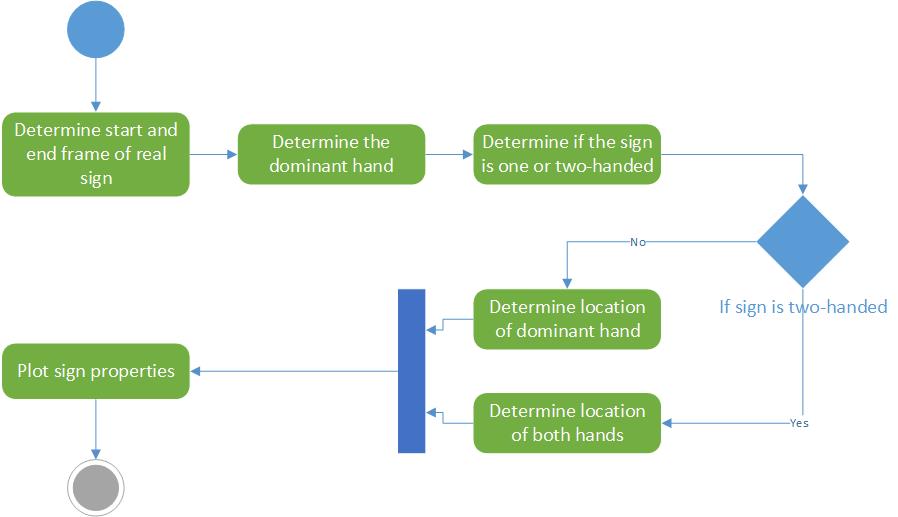


Figure 10:Sign analysis

Figure 10 shows the flow activity during analysis of the segmented data. First is determined the exact start and end frame of the real sign (the explanation what exactly is the real sign is given in section 2.4). By comparing hand displacement of left and right hand, can be determined the dominant hand and if the sign is one or two-handed. Based on this knowledge hand location can be examined (this is explained in more details in section 3.11).

## Input data

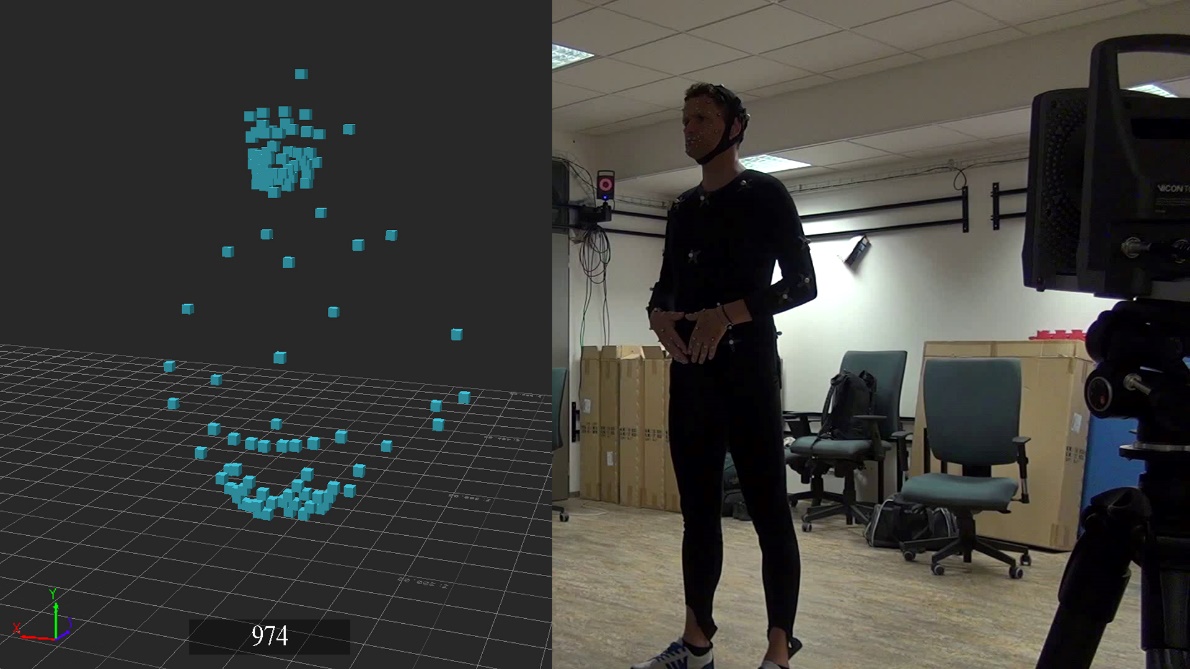
The input data is a so called dictionary file, it consists of sequence of lexical items, followed by a rest-pose (see Figure 11). A lexical item may mrefer to a singel sign or composition of two or moresigns.

Figure 11: Rest-pose: on the left MoCap data; on the right - real signer during recording

The input data is direct output of a MoCap system in .c3d file format. Basically it is a “cloud” of points with labels and coordinates for each frame. The information that needs to be extracted from this file is the 3D information about markers position in each frame, the set of used labels and the frame rate at which the recording was performed.

## Marker set

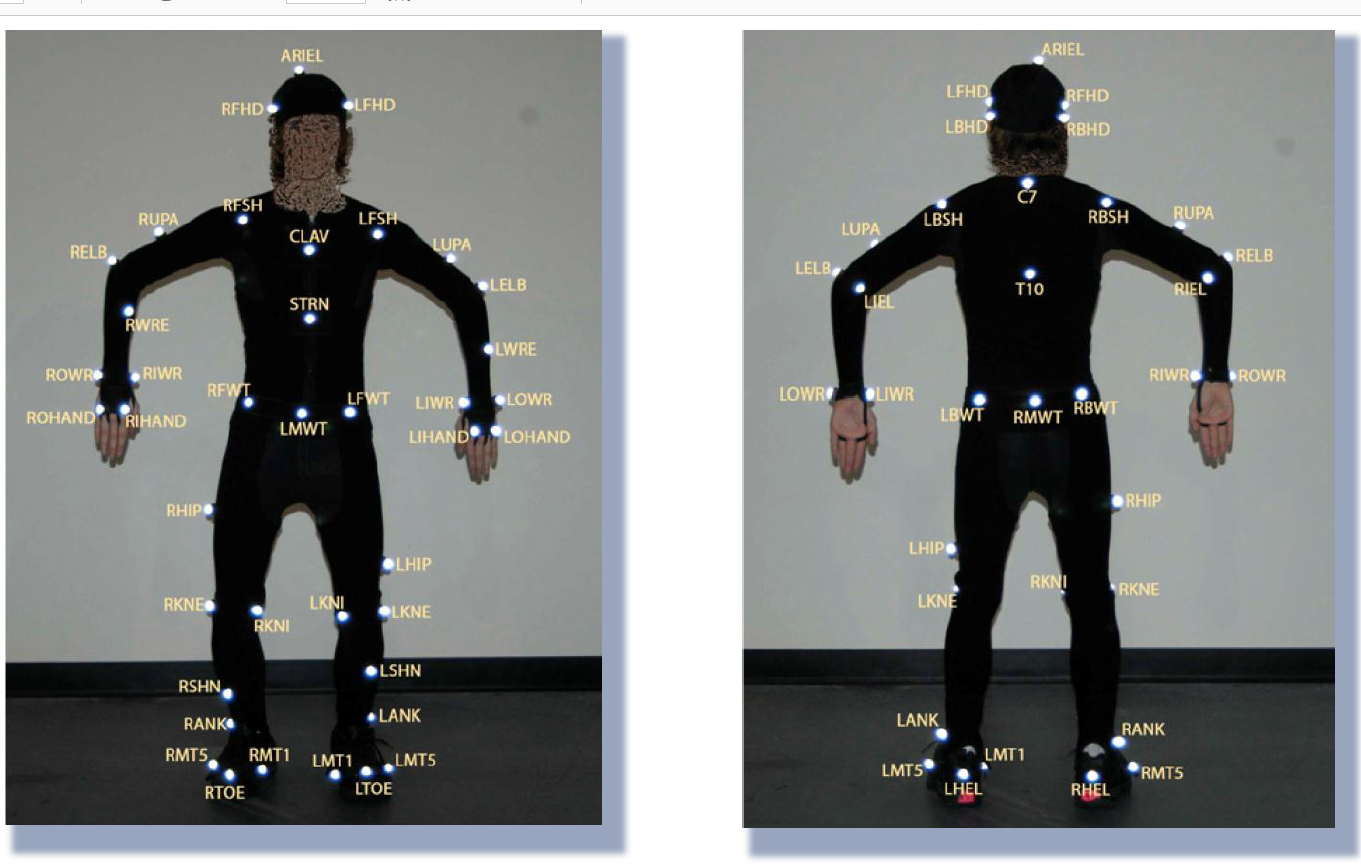
The marker set consist of 109 markers, placed over the torso, hands, fingers, head and face. The focus in this paper is in hands movement, not the movement of each marker. And trajectories of the four markers on top of hand can be interpreted as hand trajectory, so called solving process. Hand trajectory is calculated for each frame as follows:

Figure 12: Hand markers

## Segmentation

Segmentation is performed in the meaning of finding start and end frames of all lexical items in a dictionary file. A common approach for motion segmentation uses kinematics features of the movement.

* **Displacement -** defines the change in position that occurs over given period of time. It is a vector. Figure 14, shows an example of hand marker trajectory over time (measured in frames) for X, Y, Z axes.



Figure 13: Hand displacement

* **Velocity –** it is the change in displacement over time. It is calculated along the three axes. For the purpose of segmentation, I am more interested in the general movement of the hand rather than the movement along each axis, therefore a normalization is made by The Frobenius norm equation over each axis for each frame.

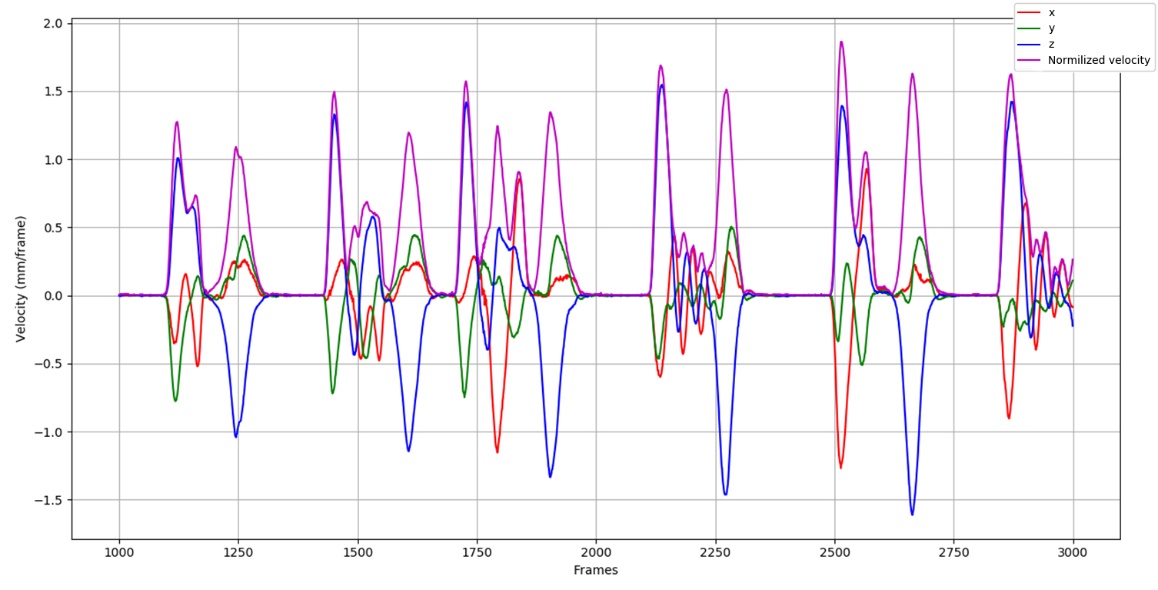


Figure 14: Hand velocity

Figure 14 shows a graph of hand velocity along the tree axes and the normalized velocity for a period of time.

* **Acceleration -** describes how the velocity changes over time.

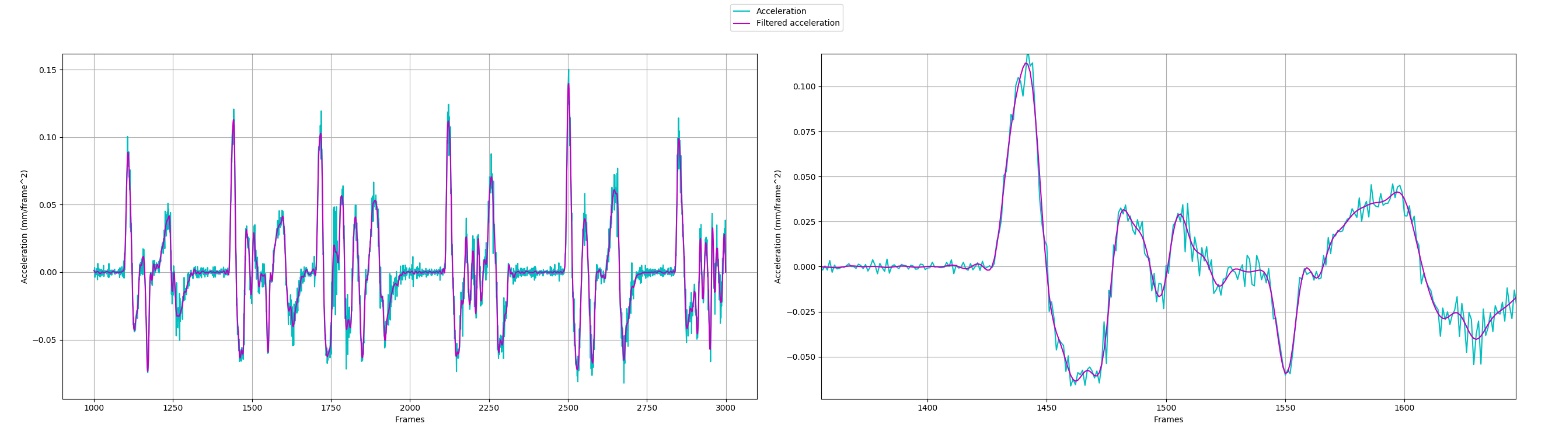


Figure 15: Hand acceleration

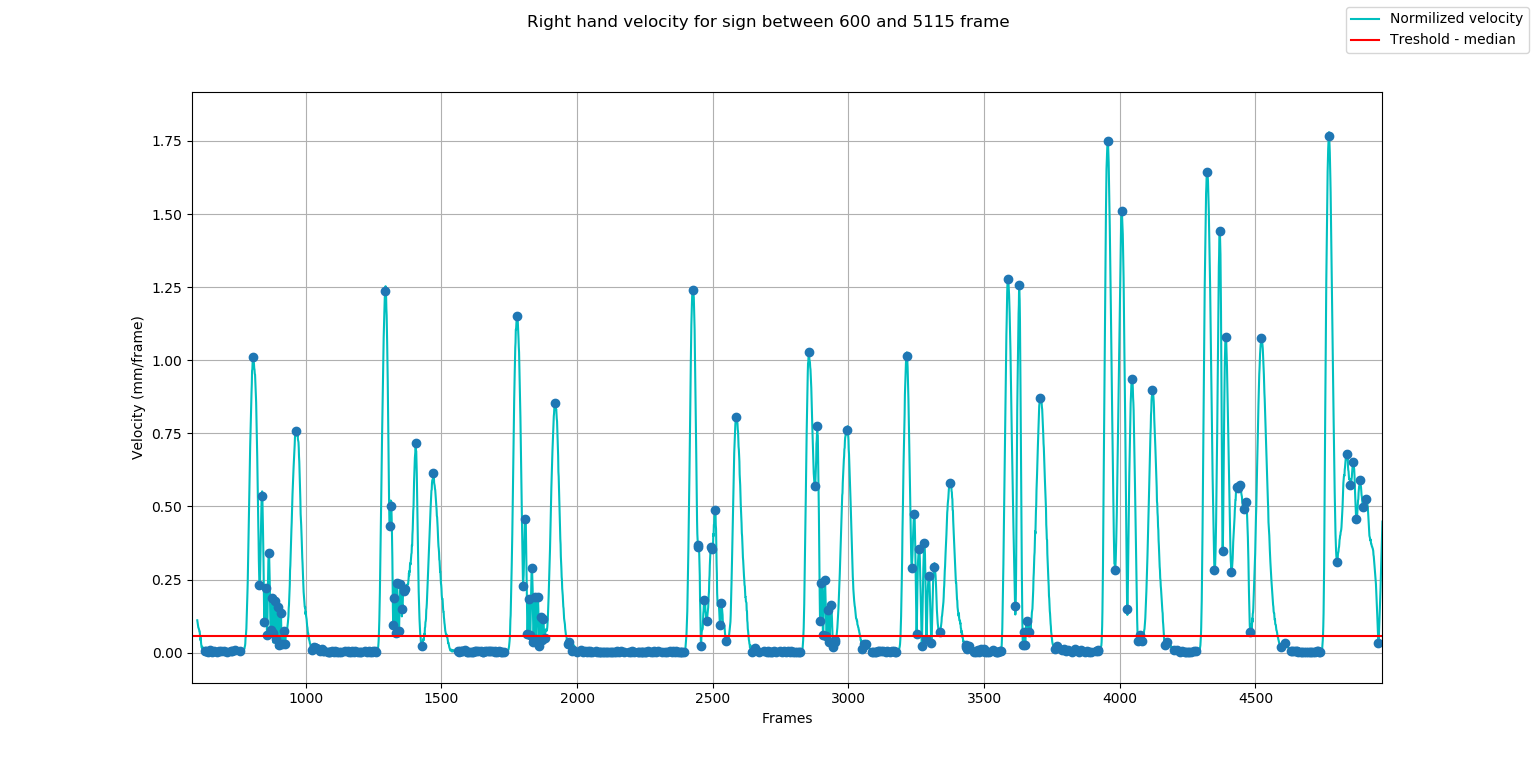
There is another aspect the has to be considered here – noise. There are many reasons for noisy data and different source of noise. Although the data presented in the time graph in Figure 13 seems quite smooth, calculating the acceleration greatly increases even the slight variations in it, as it can be seen on Figure 15 (the blue line). Butterworth filter is applied. The issue of data noise and filtering is described in more details in section 3.10.

The idea behind kinematic approach is finding the zero-crossings points in acceleration to identify significant changes in velocity. It is possible to find points of interest based on velocity extremums. Through experiments I found it more useful to base them on zero-crossing in acceleration.

The task of finding this points can be narrowed down to finding the points where the function changes its sign, meaning at which frame there are extremums in hand velocity

Next step is to decide which points are really releated with the beginning and end of the lexical item. First they are filtered by threshold function. The threshold value is chosen as median value of velocities. It gives an safe value, not to low to miss some point, neither too hight to have too many points. Figure 16 shows an example of hand velocity during movement in a whole dictionary

file, marked points of interest and the chosen theshold.

Figure 16:Hand velocity and points of interest

At this graph the correlation between hand kinematics and points of interested is easeally obesrvable.

* **Labeling -** Frames at which velocity value is unther the threshold are labeled with 0, others with 1. Then this labels are examin as vector of zeros and ones. The idea is that if the start of a sign is at certan frame it will be precceeded by at least one zero and followed by at least one 1 (ex. 0011), respectively if it is end, it would be precceded by one and fowolled by zero (1100). Ofcourse there might be some false detections, becouse of the bad threshold value or sudden halts and abrupt changes. The aim at this step is to determaine frames where at least one of the hands leaves rest-pose or where enters it. So a check of hands positoin can be performed in order to precise the labeling and reduce the errors.

Final result is pairs of frames pointing to beginning and ending of each lexical item, or frames where RP ends and next one begins. Knowing frames in which hands might be in RP, means that their beaviour during a well defined and almost static position, can be analyzed and the threshold can be refined.

## Co-ordinate system

Before recording anything with motion capture system, the systems needs to be calibrated. At this process a co-ordinate system is set. Usually the absolute co-ordinate system is on the floor in the middle of the captured volume (as in the case of recordings for SL database). This world origin may be at differ, but it has to be set before the recording.

Before any processing of the data by this tool, a relative co-ordinate system is introduced for several reasons.

SL project is based on data-driven method, so it needs a rich and reliable set of data. It should consist performances of different signers. Meaning that further analysis should not depend of person’s physique, like height weight, etc.

Also, most of the sign’s features are reviewed according to signer’s body plane, not according its global position.

For these reasons a method for changing the origin point was implemented. Considering all things said the new origin point is fixed in the middle of pelvis.

## Noise and methods for filtering

Although motion capture systems are improving over the time and tends to provide precise and accurate data, there is still the issue with noisy data. It is one of the most significant challenges that researches have been studied. In general, under the term “noise“ is understood the unwanted modifications of the motion data. There might be several sources of noise.

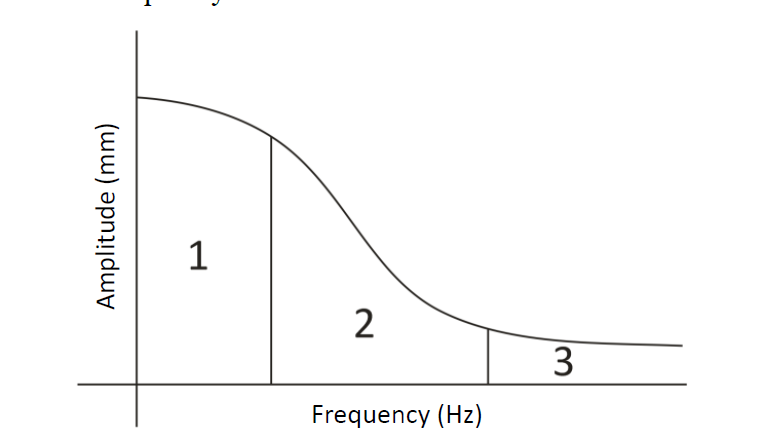
The nature of the noise and its linguistic relevance is reviewed by McDonald et al in their study for a prosodic analysis of a motion [[4](#_J._McDonald,_R.)].They analyze trajectories of different markers and use Fourier transformation to get signal’s spectrum in frequency domain. The frequency spectrum is divided into three sections which have different impact over sign analysis. According to their study based on medicals researches oscillations with frequencies higher than 10-12Hz (section 3 in figure 17) cannot be produced by human muscles thus these frequencies can be considered as noise caused by recording technology or other external factors and can be safely eliminated with low-pass filter. Other frequencies may be produced by human motion, but not all of them have meaning for sign language. The cutoff frequency between section “1” and “2” may vary according to the type of motion. Human body may produce different types of motion. For example, fingers, because of their lower mass and smaller range of movements are capable of higher frequencies than hips. They also present a table of body parts and their cutoff frequencies. Another important finding in his research is that although frequencies in the section marked with ‘2” have no linguistic meaning, they are produced by human motion and therefore its important from realistic point of view. Further analysis of these frequencies can contribute to achieving more human-like animation.

Figure 17: Division of frequency domain

A crucial step in the process of segmentation is filtering the signal. For this purpose, a Python implementation of Butterworth filter was used. It is a type of signal processing filter, based on Fourier transform. Through experimentation and based on results from McDonald’s research, the cutoff frequency parameter of 12Hz was chosen.

The following figures shows acceleration in time domain and the resulting graph after applying the filter with different cutoff frequencies.

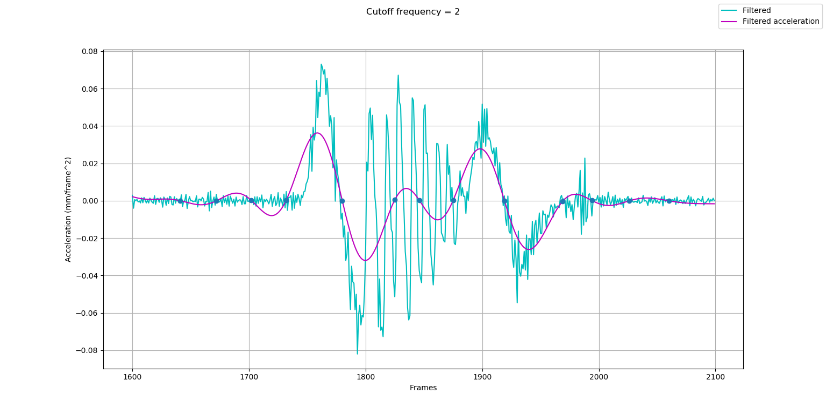


Figure 18: Cutoff frequency = 2

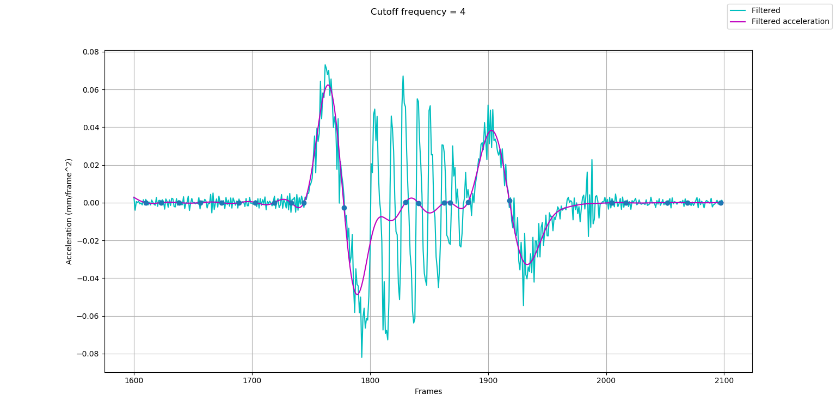


Figure 19: : Cutoff frequency = 4

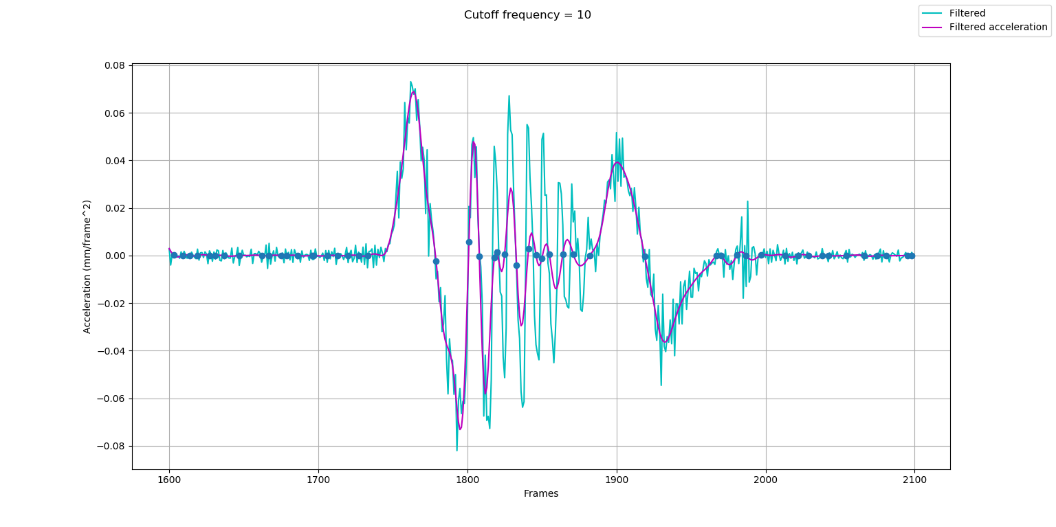
As discussed earlier according to McDonald’s research wrist limit frequency is 2Hz and fingers limit is never more than 4Hz. Compared to the results after filtering with cutoff frequency equal to 10Hz or 12Hz, I found the first two cases unsatisfying for the raw segmentation. Cutting the frequencies higher than this in section “1” cause too much loss of data and therefore may result in false segmentation.

Figure 20: Cutoff frequency = 10



Figure 21: Cutoff frequency = 12

## Sign properties

Each lexical item is investigated in the meaning of HamNoSys, which was discussed earlier in section 2.2.1. The aim of this module is to extract properties of a sign, which describe it from motion point of view. The idea is these properties later to be used for sign classification.

The dominant hand can be easily determined by computing the total length of trajectory for both hands and compare the two measurements. The idea is that no matter if the sign is performed by one hand or both, the dominant hand’s length of the trajectory will be longer. Total distance of each hand can also be used to determine if the sign is one or two-handed, by simply comparing both measurements to threshold. The threshold is dynamically computed over the general movement. If ????

### Hand Location

Hand location is determined for each frame relatively to the signer’s body plane. The space in front of the body is divided in regions according to HamNoSys (Figure 22 shows an example of body plane sectioning). The exact boundaries of each region are defined dynamically for each frame, by the position of body markers. During the sign, hands can have different handshapes, so in order to be more accurate, each region has margin, which is also calculated for each frame. The method also keeps statistical information about the number of changes in hand location and time spent in each region.

Figure 22: Example of hand location regions

### Hand Orientation

The task of defining Wrist and palm orientation can be narrowed down to minor tasks in linear algebra. The idea is to representing body parts as 3-dimensional vectors in a relative co-ordinate system.

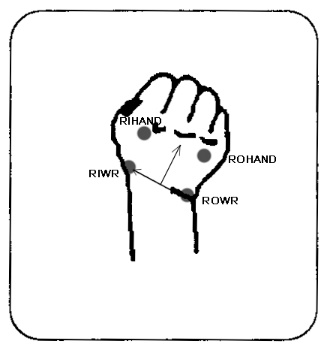
According to HamNoSys wrist orientation is notated if the wrist is bend toward the pulse or back of the hand, toward the thumb or the little finger. The task of defining the wrist bending can be solved by finding the angle between two vectors. One vector is defined by wrist markers (e.g. RIWR and ROWR), the other is defined by the middle point of segment defined of the two wrist markers and the middle point of the segment defined by the two hand markers (e.g. LIHAND and LOHAND).

Figure 23: Vectors for wrist orientation

The second mentioned vector is important, because this vector also describes the Extended Finger orientation. As it was explained earlier (see section 2.2.1.2) Palm orientation is two-step process. First has to be determined in which of the two Basic Position is the hand. This can be achieved by determining the Extended Finger orientation in the space according to the body plane.

As we know from linear algebra, a plane can be described by three non-collinear points. For example, body plane can be defined by LFSH, RFSH (markers on left and right shoulders) and one of the front waist markers (e.g. RFWT or LFWT). Therefore, two wrist markers and two hand markers are enough to define the palm plane in the space. This is useful when identifying Palm orientation.

### Handshape

Different hand shapes were discussed in section 2.2.1.1. Handshape aims to describe the complex configuration of the hand and all fingers.

At this point Handshape and Hand orientation modules are not included in the system output. They still need refinements. The description of hand orientation is tightly related to the specific handshape. A minor problem that has to be investigated more deeply is

Furthermore, during the sign handshape and orientation may change, thus leading to the question at which exact moment the handshape and orientation should be investigated.

## SVM

Support vector machines (SVMs) are powerful and flexible class of supervised algorithms for classification. Supervised means that it needs data set which has been labeled in order to train a model. The model hen can be used to cluster signs for further processing. This module is still not implemented in the tool, but it is included in the future work

# Testing

In this project testing was performed in order to prove validity and the precision of the methods used for analyzing input data. For this purpose, four dictionary files were used. Dictionary files contain recorded data on a weather thematic. Each file consists of number of lexical items.

Before proceeding to actual test, manual segmentation of the test files was performed. Results of the automatic segmentation are compared to those of the manual.

In the activity flow of the system two types of segmentation are performed. The first one is so-called “raw-segmentation” – determines two frames. *Start frame* is the moment when hand leaves the rest-pose. *End frame* is the one when hand enters RP. Second type of segmentation defines exact *Start* and *End* frame when the real sign begins/ ends.

Raw segmentation

Start frame from automatic segmentation is evaluated as successful if it is between end frame of previous sign and start frame of the current sign, defined by the manual segmentation. Respectively, end frame determined of automatic segmentation should be between end frame of the current sign and start frame of the next sign, determined by the manual segmentation. The results show 100% success of defining start frame and 85.8% of the end of the sign.

When defining the start frame manually, it is annotated the frame where hand starts the actual motion from a still position, meaning visible change in the trajectory. But human body is a complex mechanism and the hand needs preparation before starting to move. The system detects the start around 13 frames earlier, which shows that it is even more sensitive in this aspect than the human eye.

The results also show that detecting the end of the sign is harder than the start. This is due to the fact that the motion is performed by a real person, there is never a stop in the motion. It is just a steady dropdown in the speed, that is way even two different people may not agree about the exact end. It also explains why the maximum error is 51 frames, while the maximum when defining the start is 26 frames. Also the average error when determining the start is 13 frames while for end is 6 frames. The recording was performed at 120 frames per second, which means that the average error for raw segmentation is less than 0.10 seconds.

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
|  | success | average error per file (frames) | | | | average error | biggest error | |
|  |  | f01 | f02 | f03 | f04 |  | frames | seconds |
| raw segmentation | | | | | | | | |
| start tag | 100.00% | 12 | 11 | 13 | 16 | 13 | 26 | 0.216666667 |
| end tag | 85.81% | 11 | 12 | 9 | 19 | 13 | 51 | 0.425 |
| real sign boundaries | | | | | | | | |
| start tag |  | 6 | 5 | 5 | 5 | 5 | 36 | 0.3 |
| end tag |  | 8 | 6 | 7 | 7 | 7 | 52 | 0.433333333 |

Figure 24: Statistical evaluation of performed test for validating the automatic segmentation

In the study of the French sign language [[13](#_L._Naert,_C.)] they use the kinematic based segmentation to refine a manually segmented data. In their study they consider the 15 frames (for a capture frequency 100Hz, corresponds to 0.15s of motion) difference between manual and automatic segmentation as relatively small and insignificant. Based on their research I found that difference of 13 frames is quite satisfying, but it has to be taken onto account the fact that this results are over relatively small amount of data and further testing is needed.

# Conclusion

Sign language is the natural language for a people with speech and hearing impairments and their firs mean of communication. Sign language vary from culture to culture and from region to region. Although it is the main alternative of spoken language for deaf community, both languages are very different from any point of view. The difference in linguistic aspect makes even the written form of spoken language difficult to understand, especially for prelingual deaf people. These barriers in communication and access to sources of information motivates many researches for SL synthesis. The aim of this works is developing tools for providing human –like movements while signing, which can be accepted from deaf community. Due to this fact and because SL utterance can be seen as continuous stream of motion, recent research based on data-driven synthesis and data from motion capture has been developed. Animations based on data from real signer is more realistic. A weak spot of this approach is that post-processing of MoCap data and skeleton reconstruction are laborious and time-consuming processes, as well as manual segmentation and annotation of SL utterance.

My work aims to develop a tool that can help in processing the data for SL database. I reviewed Hamburg notation system as a way to understand SL from phonetic point of view. My work is focused on hand movement and its kinematic features such as position and its derivatives (velocity, acceleration) in order to decompose sign and extract properties than describe the sign.

First step is implementing an algorithm for segmentation of SL dictionary file. The aim is identifying starting and ending moment of a lexical item. The algorithm is based on analysis of hand trajectory and velocity. Unlike speech in which transitions between words is silence, movement never stops. However, there is phase where hand prepares for execution of the sign, and this phase may be indicated by drops in velocity. Values are processed with low-pass filter to prevent the algorithm from detecting incidental minima due to noise in data.

After proper segmenting of the data and extracting only the motion that will have semantic meaning, is possible to proceed with further analysis. At next steps the meaningful parts of the sign are studied. The algorithm aims to extract sign features such as Hand location, Hand orientation and dominant hand.

The next step is to extended the number of sign features extracted from the motion analysis and to utilize computer learning methods to cluster signs for further processing.

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