

The AutoActive Research Environment

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Software

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Statement of Need

There is an ever-growing variety of biomedical sensors and wearables that aim to monitor activity, biomarkers, and vital signs. However, to fully understand the physical and physiological factors of the underlying processes, multiple sensors are often needed in combination with videos. Software for combining, synchronizing, organising and processing sensor data from multiple sensors and videos is therefore essential. Even though multiple open-source solutions like Pyomeca ([Martinez et al., 2020](#)) and ALPS ([Musmann et al., 2020](#)) exist, existing software solutions are limited. None provide the possibility to combine sensor data and videos, few provide tools for synchronising sensors, and none provide tools for synchronising sensors with videos. Furthermore, many solutions rely on cloud storage, which is often unacceptable in biomedical research. To meet these limitations, we have developed the AutoActive Research Environment (ARE). The idea of ARE is to create a generic methodological framework, supporting a wide range of sensors and tools that aid the development, optimisation, and evaluation of algorithms to extract high-level, quantified analysis from the sensor data.

Summary

ARE consist of three different open-source software modules; ActivityPresenter and the accompanying Matlab and Python toolboxes. ActivityPresenter is created to simplify the process of visualising, synchronising, and organising data, such as sensor data and videos from multiple sources, while the Matlab and Python toolboxes allow researchers to easily process data. Furthermore, a file format called AutoActiveZip (aaz) was created to store data and metadata in an organized manner. This format is a structured archive which contains immutable data structures and where the information within can be accessed without the use of temporary files that needs to be cleaned up. This ensures that sensitive data is not inadvertently left in temporary folders in case of program failure. The format allows the strengths of ActivityPresenter, such as synchronising data from multiple sources, and visualising videos and sensor data side by side to be combined with algorithms developed in Matlab and Python.

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Overview of the AutoActive Research Environment

ActivityPresenter

ActivityPresenter is an easy-to-use software with a graphical user interface that can visualise, synchronise, and organise data from sensors and cameras. It is developed using the Xamarin framework which simplifies the task of supporting multiple operating systems, as almost all the code for both the GUI and data handling can be shared for all targeted platforms. An example of ActivityPresenter can be seen in Figure 1 where we visualise synchronised video with gyroscope and heart rate data.

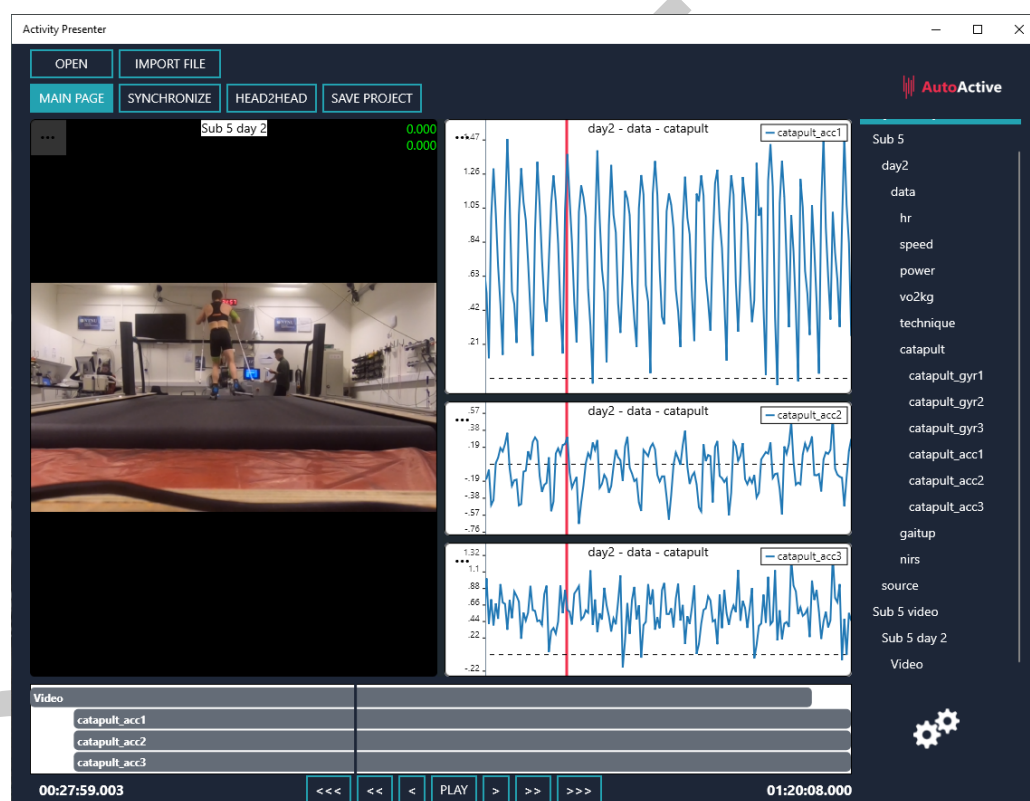


Figure 1: Example of ActivityPresenter.

ActivityPresenter makes it possible to import data from an aaz file, files generated by sensors such as the Physiolog 5 sensor (Gait Up SA, Lausanne, Switzerland), Catapult Optimeye S5 (Catapult Sports, Melbourne, Australia) and tcx files from Garmin (Garmin International Inc, Kansas, USA) as well as common data types such as csv and xlsx and all video formats acceptable by Microsoft Windows. Furthermore, it is easy to extend accepted file types as the architecture is plugin-based. The application architecture can be seen in Figure 2. Data can be loaded from an aaz file or from a custom data importer onto the databus. Data on the databus can be used to create tree views to get an overview of the datasets, create figures to view data, or synchronise data coming from different sources. When the data has been synchronised, the changes are written back to the databus. The data on the databus can also be reorganised and written to a new aaz file. When data is made available on the databus, data is not necessarily loaded into memory. If data comes from an aaz file, data is first read into memory when accessed. This typically happens when the data is selected from the data tree view for visualization. Hereby, we minimise the memory footprint of ActivityPresenter and make it possible to view parts of large sessions on devices with limited resources.

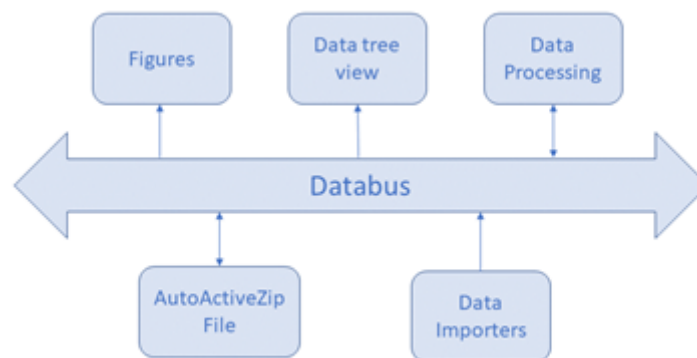


Figure 2: Architecture of ActivityPresenter.

Data Handling

Session

The concept of sessions is key to how we handle and store data in ARE, they are the root containers of datasets. A session represents an activity bounded in time and space and stores the information about the context of the activity, and the data generated during that activity. When a session is saved it becomes immutable and is assigned a unique identifier. This allow sessions to be based on previous sessions, and enables traceability and reproducibility as analysed information is referenced to the session where the data for that analysis was stored. It also allows referencing of large files, such as high-quality video captures during an activity, without duplicating the files.

AutoActiveZip File

An AutoActiveZip (aaz) file is an uncompressed zip archive that contains a set of sessions. All information inside the archive is organized into directories like a file system. By using the ZIP archive file format, we can store multiple datasets in a single file, store necessary metadata to describe the data, but most importantly we can compress data in a binary format suitable for different data types. In ARE sensor data is stored in parquet files, while videos are stored in their original format, as they are already compressed. By using the parquet format instead of a common data format such as csv we reduce storage requirements on larger files. The necessary metadata is stored in a json file within the archive. An example of the structure of an aaz-file can be seen in [Figure 3](#) where a file named “GoPro_openpifpaf.aaz” is described. An aaz file can contain one to many sessions, in this case there is only one session. The session is identified with a uuid4 number, “a9ee0260-c8c6-4f87-b00c-bb25a7772885” and is just a folder. Inside the session folder two more folders exist, a folder named “data” and a folder named “video,” and a json file named “AUTOACTIVE_SESSION.json.” The purpose of the json file is to store necessary metadata. The folder named data contains a parquet file named “video_features.3ded455b-6567-440f-8f87-effc6549ed05.” The other folder contains a video named “GoPro1_inne.d84020b3-f047-4026-96df-51ce987d747e.” Although the folders only contain one file in the example, a folder can potentially contain multiple different files.

- GoPro_openpifpaf.aaz
 - a9ee0260-c8c6-4f87-b00c-bb25a7772885
 - data
 - video_features.3ded455b-6567-440f-8f87-efc6549ed05
 - video
 - GoPro1_inne.d84020b3-f047-4026-96df-51ce987d747e
 - AUTOACTIVE_SESSION.json

Figure 3: Structure of the AutoActive Zip file

85 Json file

86 Each session in an aaz file contains a root json file which contains important metadata used
 87 when reading data from the aaz file into native C#, Matlab and Python classes. An example
 88 of the structure of a json file can be seen in Figure 4. The json file corresponds to the zip file
 89 in Figure 3. Each part of the json file contains a meta field and a user field. The meta field
 90 contains only data not presented to the user, while the user field contains metadata visible
 91 and editable by user or nested datasets in the form of other data objects. The first meta
 92 field seen in Figure 4 describes the session object while the user field contains another folder
 93 named "Data" where the data-table and video is defined.

```

1 {
2   "meta": {
3     ...
4   },
5   "user": {
6     "created": "2021-05-21T17:46:21.861+01:00",
7     "name": "Openpifpaf with deepsort",
8     "Data": {
9       "meta": {
10        ...
11      },
12      "user": {
13        "video_features": {
14          "user": {},
15          "meta": {
16            "type": "no.sintef.table",
17            "version": 1,
18            "attachments": [
19              "/data/video_features.3ded455b-6567-440f-8f87-efc6549ed05"
20            ],
21            "units": {
22              ...
23            },
24            "is_world_clock": false
25          }
26        },
27        "video": {
28          "meta": {
29            ...
30            "attachments": [
31              "/video/GoPro1_inne.d84020b3-f047-4026-96df-51ce987d747e"
32            ]
33          },
34          "user": {
35            "file_name_full": "C:\\...\\d41da579-5cd1-40ef-bb27-a125777eeb16\\videos\\GoPro1_inne.d84020b3-f047-4026-96df-51ce987d747e",
36            "file_details": {
37              "meta": {
38                "type": "no.sintef.folder",
39                "version": 1
40              },
41              "user": {
42                ...
43              }
44            }
45          }
46        }
47      }
48    }
49  }
50 }
51 )

```

Figure 4: Structure of the json file inside the aaz file

94 Matlab and Python toolboxes

95 The Matlab and Python toolboxes consist of a reader and writer class which reads and writes
 96 AutoActiveZip files. The Matlab toolbox also consists of classes which makes it easy to
 97 create aaz dataobjects directly from e.g. Physiolog 5 and tcx files from Garmin. Furthermore,
 98 the libraries consist of a set of classes storing sensor data and videos. The data classes all

99 inherit from the dataobject class. The dataobject class specifies all transformations needed
100 for converting the native Matlab and Python formats to and from the AutoActiveZip file. All
101 dataobject sub-classes in Matlab and Python are identified by a type string. The type string
102 is also stored in the AutoActiveZip File in the json file and can be seen in the meta sections in
103 [Figure 4](#). This makes it possible to load a session from the AutoActiveZip into specific data
104 classes.

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