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³ HEP Software Foundation Community White ⁴ Paper Working Group – Visualization

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³⁰ ABSTRACT: In modern High Energy Physics (HEP) experiments visualization of ex-
³¹ perimental data has a key role in many activities and tasks across the whole data
³² chain: from detector development to monitoring, from event generation to recon-
³³ struction of physics objects, from detector simulation to data analysis, and all the
³⁴ way to outreach and education. In this paper the definition, status, and evolution
³⁵ of data visualization for HEP experiments will be presented. Suggestions for the
³⁶ upgrade of data visualization tools and techniques in current experiments will be
³⁷ outlined, along with guidelines for future experiments. This paper expands on the
³⁸ summary content published in the HSF *Roadmap* Community White Paper [1].

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71 **1 Scope**

Visualization: Turning numbers
into pixels

72

Hadrien Grasland
HSF Workshop 2017, Annecy

73 This paper will describe three kinds of data visualization used in High-Energy
74 Physics (HEP): interactive visualization of event data in applications known com-
75 monly as event displays, statistical data visualization such as histograms, and non-
76 spatial data visualization such as networks and graphs.

77 Event displays are the main tool used to explore experimental data at the event
78 level. There are two main types of displays. The first type are those that are
79 integrated into an experiment’s software frameworks, which are usually able to access
80 and visualize all experimental data at the cost of greater application complexity and
81 lesser portability. The second type of displays are those designed as cross-platform
82 applications, lightweight and fast, often delivering a simplified version or a subset of
83 the event data. All event displays show the detector geometry; the level of detail
84 displayed depends on the application’s use-case and targeted audience as well as on
85 the application’s capability to render geometries responsively.

86 Beyond event displays, HEP also uses statistical data visualizations such as his-
87 tograms, which display the distributions of data variables in aggregate over multiple
88 events. These visualizations are not strongly linked to a detector geometry. Data
89 analysis tools and techniques used in HEP are described in the HSF *Data Analysis*
90 and *Interpretation* Community White Paper [2].

91 The final types of visualization considered in this paper are those that visualize
92 non-spatial data, such as the graphs used to visually describe the structure of the de-
93 tector description, that is the representation of all geometrical volumes that compose
94 the sub-detectors and the infrastructure of a HEP experiment. More details about
95 the detector geometry can be found in the HSF *Detector Simulation* Community
96 White Paper [3].

97 Other types of data visualization used in HEP experiments, such as visualization
98 for slow control or dashboards for data analytics, are considered out of scope and are
99 not discussed.

100 The content of this paper is the summary of the direct experience of the authors
101 in designing, building, and deploying interactive data visualization applications for
102 the experiments ALICE, ATLAS, Belle II, CMS, LHCb, and LSST, for scientific
103 software frameworks such as ROOT, and in other cross-experiment projects. It is
104 the outcome of a number of discussions and workshops organized under the auspices
105 of the HEP Software Foundation, in which the authors worked to build a common,

106 shared view of the field, its current issues, and the potential ways it could and should
107 evolve in the context of HEP.

108 2 Current landscape

109 2.1 Event displays

110 Three key features characterize HEP event displays. The first is an *event-based workflow*. Applications access experimental data on an event-by-event basis, visualizing
111 the data collections belonging to a particular event. Data can be related to the actual
112 physics events (*e.g.* a collection of reconstructed physics objects, like jets and tracks)
113 or to the experimental conditions (*e.g.* different versions of the detector description
114 and calibration data).

116 The second key feature is *geometry visualization*. The level of geometric detail
117 displayed depends on the specific use-case, on the way the geometry information is
118 stored and fetched (*e.g.* from a database as part of a software framework or from an
119 external file), and on limitations of the application itself along with considerations
120 about speed, efficiency, and portability.

121 The third key feature is *interactivity*. Applications offer different interfaces and
122 tools for users to interact with the visualization, select event data, and to set cuts
123 on objects' properties. In addition to the interactive usage, applications often store
124 different settings to automate user actions.

125 In the following subsections several important aspects of data access, application
126 development and distribution, and geometry description and visualization, as they
127 pertain to the current landscape of event displays, are discussed in more detail.
128 Screenshots of several event displays can be seen in Figure 1.

129 2.1.1 Data access

130 Access to event data comes either natively or via intermediate formats. In the former
131 case direct access of native event formats is only possible for an application integrated
132 with the experimental software framework.

133 There are several advantages to having access to the experiment's framework,
134 such as full access to the experimental data in its native format and to software tools,
135 services, and databases. Through them, event display applications can make use of
136 the full detector simulation geometry, of conditions data, and of all the framework's
137 application program interface (API).

138 One disadvantage of this approach is that full support for the display application
139 is often limited to those platforms on which the framework itself is supported, limiting
140 cross-platform distribution and support. One way to mitigate this is to distribute
141 a light version of the framework along with the application; CMS Fireworks [10]
142 takes this approach. However, issues of platform support for the light framework

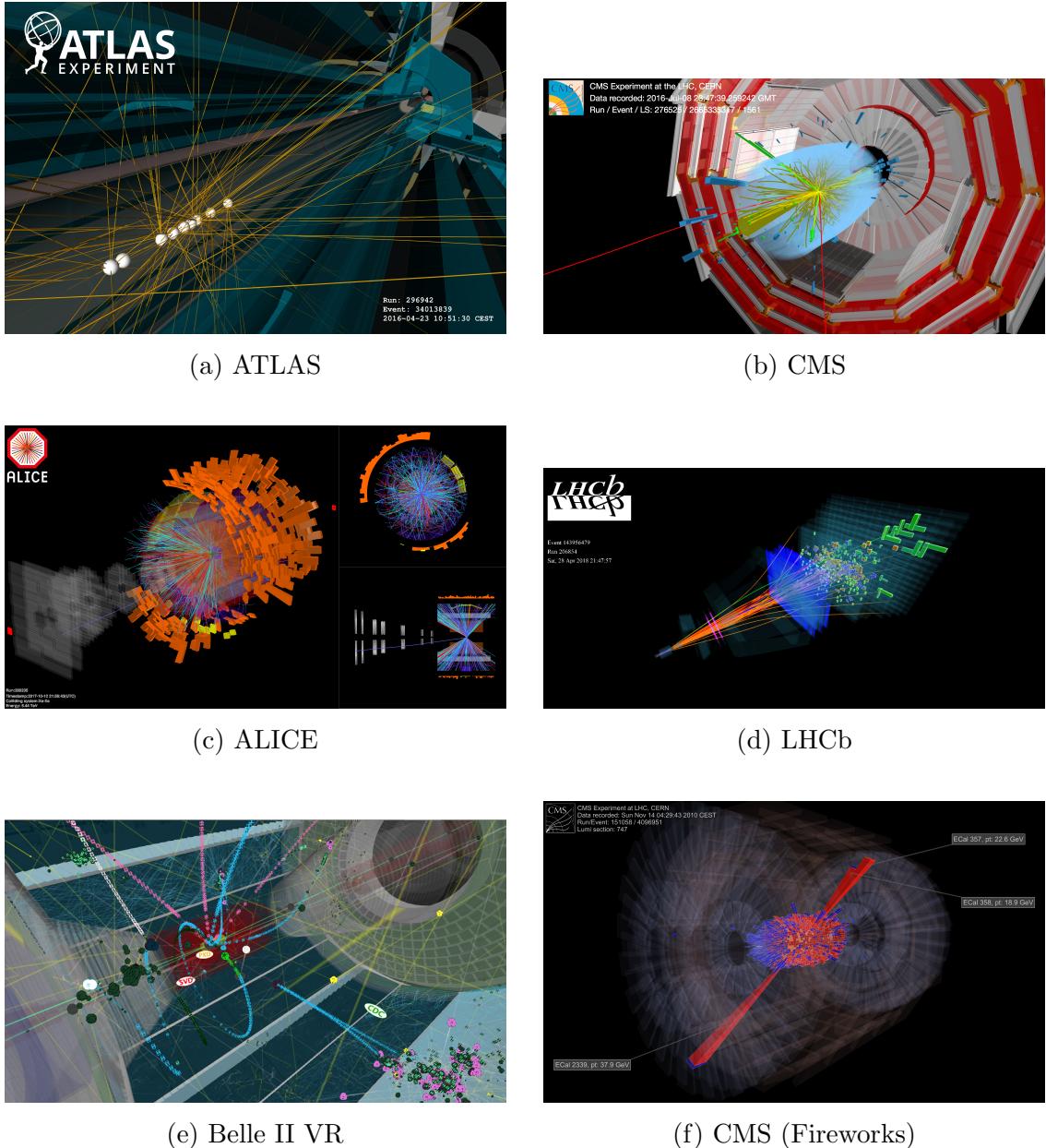


Figure 1: Different examples of HEP 3D event display applications: **a)** an ATLAS 3D event display made with VP1 [4, 5]; **b)** a CMS 3D event display made with iSpy [6]; **c)** an ALICE 3D event display [7]; **d)** an LHCb 3D event display [8]; **e)** a Belle II VR event display [9]; **f)** a CMS 3D event display made with Fireworks [10]

and for the visualization application can still exist. A further disadvantage to the full-framework (and even light-framework) approach is that one must also support various versions of the data format along with the underlying framework API. In addition, users have to have knowledge of the framework in order to interactively explore and visualize event-based data. Lastly often the user-interface to a full-

148 framework application is geared towards the expert.

149 The latter approach to data access is via an intermediate format. Usually the
150 data needed for visualization is a subset of the full information found in the native
151 experimental format. Therefore one can extract what is needed from the framework
152 through the usage of dedicated exporting software tools and store in intermediate
153 formats.

154 With the use of an intermediate data format (usually based on different flavors of
155 the XML or JSON formats) the event display application is potentially separate from
156 the experimental software framework and therefore its deployment is not limited to
157 those platforms officially supported by the experiment. It is then possible to have
158 both lightweight data and applications, which can be easily and broadly distributed.

159 The primary drawback to this approach is that, with no direct access to the
160 experimental data, some information is necessarily not accessible. Moreover, every
161 time there is the need to modify the content of the intermediate data files, one
162 needs to run the data extraction tools on the native data again. In addition to that,
163 only events which are identified as potentially interesting are extracted and their
164 data reduced to be stored in the intermediate data file; so, if the end user wants to
165 analyze and visualize other potentially interesting events, the extraction/reduction
166 step must be performed again on the relevant data.

167 Regardless of the approach to data access one must consider the use-case: what is
168 useful or necessary in one use-case may not be for another. A graphical representation
169 of the two approaches can be seen in Figure 2.

170 2.1.2 Application development and distribution

171 Currently, the two most common ways of distributing event display applications are
172 as a desktop application and as a web application running in the browser. Each
173 approach has its advantages and disadvantages which are further described in this
174 section. The current landscape is summarized in Figure 4 and is further described
175 in this section.

176 Native mobile applications running on devices such as smartphones and virtual
177 reality applications are less common in HEP. However, they are a growing feature in
178 the current landscape and many experiments are exploring the possibilities of those
179 emerging technologies. At the end of this section we describe briefly the mobile
180 applications released so far. Further developments will be described in Section 3.4.

181 **Desktop applications** Many experiments have developed integrated event-
182 display applications in C++, which is the main language used for developing HEP
183 software frameworks, on top of the OpenGL [11] APIs. The choice of the OpenGL
184 API, compared to other APIs like Direct3D [12], resides in its cross-platform nature
185 as OpenGL is an open standard. The OpenGL consortium defines the API: the inter-
186 face with which all the implementations have to comply. The actual implementation
187 is provided by vendors, usually targeting a specific hardware. Many hardware and

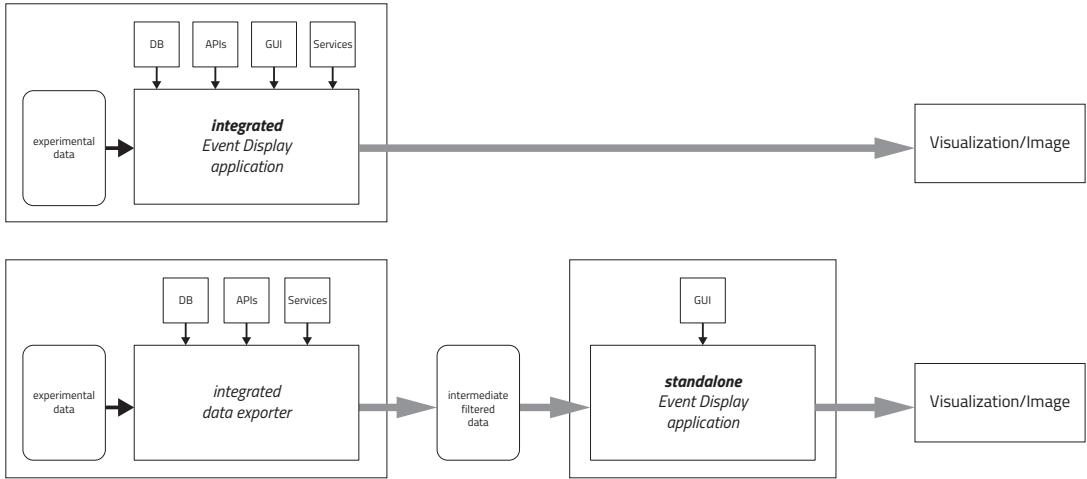


Figure 2: The two different types of event display applications. At the top, the framework-integrated event display application is able to access all experimental data, all services, APIs, and databases provided by the experiment’s framework; as a drawback, the application must be run on specific platforms supported by the framework and it must use only graphics and GUI libraries compatible with them. At the bottom, the standalone approach, where experimental data are accessed, filtered, and extracted by using custom data exporters, which create intermediate data files containing only the interesting pieces of information; then, a standalone event display application reads those data in and it creates the required visualization; the advantage is having a cross-platform application which can use any graphics and GUI libraries, while the drawback is the lack of direct and full access to the experimental data and to the experiment’s software tools, which prevents a detailed, full visualization.

188 software companies such as Intel and NVIDIA are part of the OpenGL consortium,
 189 which assure the support and the lifetime of the OpenGL API.

190 Some HEP visualization applications use OpenGL calls directly through custom
 191 graphics engines. This is the most robust approach as the developers can take full
 192 control over the OpenGL interface and the project can be independent of other
 193 software libraries. Two examples of HEP applications which follow this path are
 194 the ATLAS Persint application [13] and the ROOT EVE toolkit [14], which is used
 195 both by the CMS Fireworks application [10] and the ALICE Event Visualization
 196 Environment (AliEve) [15]. One disadvantage of this approach is that developer
 197 time has to be assured to maintain the graphics engine itself, on top of the effort
 198 needed for the development and maintenance of the event display application.

199 Other applications use higher-level interface libraries as graphics engines. This
 200 has the advantage of delegating a large part of the lower-level development work to

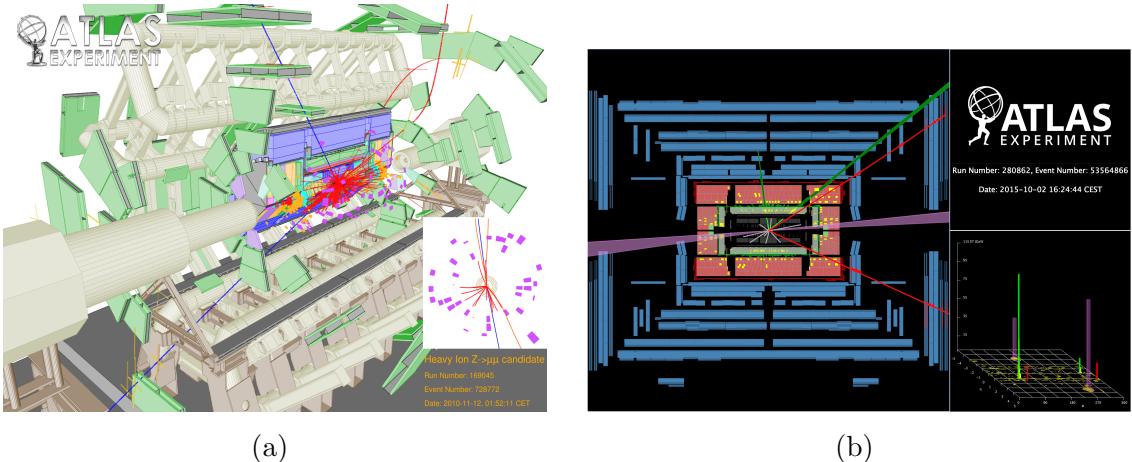


Figure 3: a) an ATLAS 3D event display made with Persint [13, 16]; b) an ATLAS 2D event display made with Atlantis [17, 18].

external software packages, leaving the developers to concentrate on the application development itself. A popular graphics library used in HEP software has been Open Inventor [19], used by the defunct CMS Iguana [20, 21] application, or its clone implementation Coin (also known as Coin3D) [22], which has been used by applications such as ATLAS VP1 [5], LHCb Panoramix [23] and the defunct desktop version of the CMS iSpy application [24]. Coin / Open Inventor was chosen because of its integrability in C++ code, its performance, and its coding style. Moreover, the way Open Inventor handles graphical volumes could be easily matched with the way geometry volumes are handled to describe the detectors in HEP experiments. Open Inventor organizes geometry volumes as a series of nodes in a tree-like structure in the same way as some HEP experiments do. ATLAS, for instance, developed their geometry library ‘‘GeoModel’’ [25] based on the same tree-like structure of nodes used by Open Inventor.

The drawback of this approach is the dependency on external software projects, which could end up with a loss of functionality if third-party library development and support are abandoned. Many scientific visualization applications, also in fields other than HEP, faced this when the support of the Coin library was dropped by the company that led its development [26]. The result is the aging of libraries which after a while show incompatibilities with modern hardware, compilers, and platforms. The time spent by HEP developers to repair or to maintain those abandoned libraries results in time not spent on actual development of the software applications themselves.

An additional approach to the development of event displays is to create and distribute an application using the Java programming language. The Atlantis [18] program and its derivative MINERVA [27], which is used as an educational tool, both developed for the ATLAS experiment, are based on the Java graphics libraries and

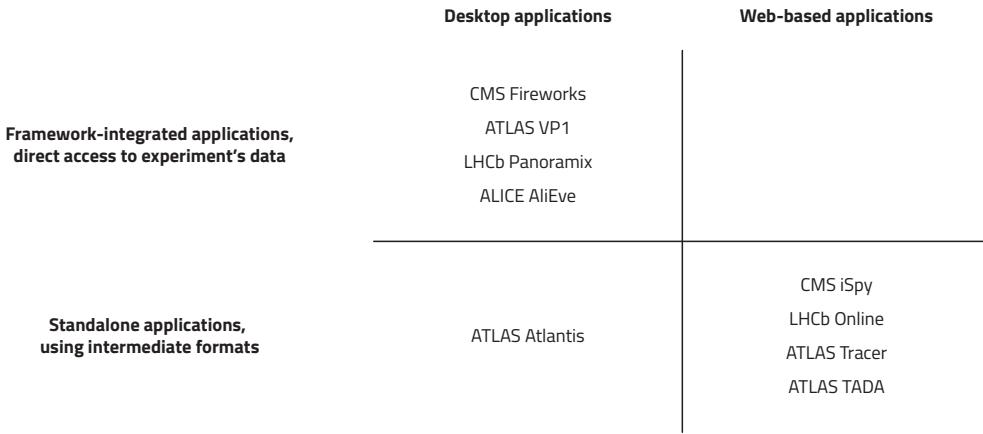


Figure 4: The image summarize the current landscape of event display applications in HEP. Many experiments developed full-framework desktop applications as well as light, web-based applications. As one can see from the plot, there are no examples, so far, of full-framework applications using web-based visualization graphics, due to the data access issues with today’s experiments’ frameworks. A new approach in that direction is what it is suggested in this paper, in Section 3.3.

227 can be run either online in a web browser or stand-alone on a desktop machine.

228 **Web-based applications** Several experiments at the LHC, notably CMS [28],
 229 LHCb [29], and ATLAS [30, 31] have created web-based event displays using We-
 230 bGL (Web Graphics Library) [32]. WebGL is a JavaScript API that conforms to
 231 OpenGL ES (a subset of the OpenGL API for embedded systems) conceived for
 232 rendering interactive 3D and 2D graphics within any compatible web browser with-
 233 out the need of external plug-ins. With WebGL in the browser one can combine
 234 high-quality graphics with the functionality and accessibility of the browser. This
 235 combination of graphics and user function was previously only available via bespoke
 236 desktop applications based on OpenGL and graphical user interface toolkits such
 237 as Qt [33]. Browser-based event displays have several distinct advantages: they are
 238 easy to distribute to the user, they can be prototyped quickly, and the client is often
 239 much lighter-weight, as the need for building, packaging and distributing external
 240 libraries is greatly reduced. There are also several mature and actively developed
 241 WebGL frameworks, such as the popular three.js [34], that provide straightforward
 242 and simplified APIs for ease of development.

243 **Mobile applications** Mobile devices such as smart phones and tablets are more
 244 and more ubiquitous and these devices are used more and more as substitutes for
 245 desktop and laptop machines. However, mobile devices still do not have the comput-
 246 ing power usually needed for HEP data analysis, where huge amount of experimental
 247 data are retrieved and processed. In addition they usually run dedicated operating

systems whose self-contained nature makes their integration within the HEP workflow difficult, particularly for the statistical-based visualization used in data analysis, described in Section 2.2. Despite the current limitations for these use-cases mobile devices can be found in the current landscape of event displays.

Several event display applications have been developed for, or at least can be run on, mobile devices. An example of a native application is LHSee [35] which live-streams ATLAS events, processed and extracted through Atlantis [18], to a user’s phone and provided contextual information on ATLAS and the events being displayed. The CMS iSpy WebGL application [28] runs on mobile devices in the browser. The Camelia application [36], developed by the CERN Media Lab using the Unity game engine [37], can be built as a mobile application by using the tools provided by the game engine and run on mobile devices as well. More details on game engines can be found in Section 3.4.1.

Virtual and augmented reality applications Virtual Reality (VR) simulates the user’s physical presence in a virtual environment, and the application is typically run on a head-mounted display that provides visual and aural experience of the simulated environment. Different degrees of realism and immersion are possible, depending on the targeted hardware.

VR technologies can be used to build immersive applications, to let the general public virtually visit HEP detectors and explore experimental sites. Many HEP experiments started developing VR applications, mostly as an educational tool, for outreach events. These include ATLASrift [38] and Belle II VR [39]. As many HEP experiments are not accessible during data taking or are classified as supervised areas due to security or safety issues, VR applications let the HEP community open their sites and experiments to the general public. Also, they let people look at simulated collisions in a simplified yet realistic environment, which help people acquire the basic concepts on which HEP experiments are built and run. Such applications are currently used in public events, in museums and science centers, and in meetings with the governments and the funding agencies.

Augmented Reality (AR), instead, uses a camera to take a view of the real world around the user and screen where simulated objects are rendered in 3D and shown on top of that image dynamically, following the user’s interaction and motion. This lets the user move within an environment where real and simulated objects live together. AR can be used in HEP as an educational tool, for instance to dynamically show and describe a HEP detector to a group of people or a class. Some HEP experiments have been started exploring AR technologies for HEP, particularly ALICE, ATLAS, and CMS.

Both VR and AR applications are usually developed in specialized graphics engines, which prevent the interaction with the experiment’s framework to access native data. More details on VR and AR applications for HEP are found in Section 3.4.3.

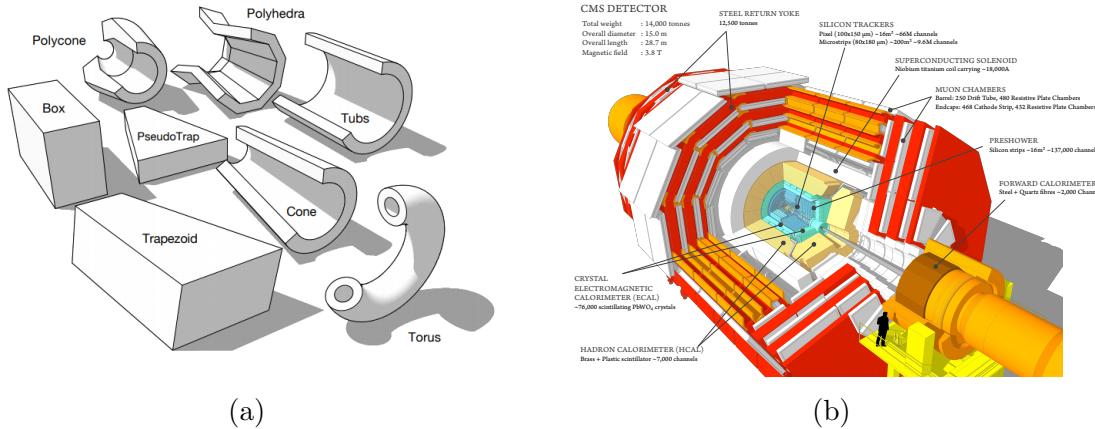


Figure 5: Use of a 3D software (SketchUp) to render the basic geometrical shapes imported from the CMS detector description and the final volumes [40].

288 2.1.3 Geometry description and visualization

289 Geometry visualization provides important visual context for event displays and dedi-
 290 cated geometry displays are useful applications by themselves. There are typically
 291 three levels of detail found in applications. The most detailed geometry is typically
 292 called the *simulation geometry* and can include the sensitive elements of the detector
 293 as well as support structure. Detector experts are typical end-users of applications
 294 that display this level of geometrical detail. Less detailed is the so-called *recon-
 295 struction geometry*, which describes the sensitive elements of the detector such as
 296 calorimeter crystals and wire and strip chambers. It is this level of detail that is
 297 typically found in event displays. The least-detailed geometry descriptions are those
 298 that are simplified versions of the detector geometries and are used to provide visual
 299 context only.

300 Currently, different geometry formats and libraries are used in HEP. Some ex-
 301 periments use their own custom format (*e.g.*, [25]), while others use the geome-
 302 try tools provided by the ROOT framework [41]. More recently, some attempts
 303 have been made to build common formats and libraries for detector geometry, like
 304 DD4HEP [42], adopted in conceptual design studies for future high-energy colliders,
 305 including the CLICdp [43] and FCC [44] collaborations. In all cases detector volumes
 306 are built from simpler geometrical entities: geometrical shapes like Tube, Cone, Box,
 307 and more complex variations of these are combined in order to build the volumes of
 308 the experiment’s geometry. Geometry libraries and formats are also described in the
 309 HSF *Detector Simulation* Community White Paper [3].

310 The differences in geometry formats used by the different experiments, by de-
 311 tector simulation programs like Geant4 [45], and by data analysis frameworks like
 312 ROOT, typically require developers of visualization applications to write converters
 313 between the different formats. In addition it is often not easy to use a visualization

314 tool developed for one experiment with another one as current visualization tools are
315 often tightly bound to the geometry library used by the experiment.

316 In framework-based applications the geometry information can come directly
317 from the experiment’s detector description and in many cases the hierarchical struc-
318 ture of the detector description is preserved and accessible. Standalone applications
319 typically use a geometry file with information exported from the software framework.
320 An example hybrid solution is the one developed for the CMS experiment using the
321 SketchUp application [46], described in Ref. [47] (see Figure 5). The CMS detector
322 description [48] as written in XML is parsed using Ruby scripts and 3D models are
323 built using the SketchUp program via its Ruby API. SketchUp can then export to
324 various standard 3D file formats. In this way detailed simulation geometry can be
325 available in a standalone application.

326 2.2 Statistical data visualization

The simple graph has brought more
information to the data analyst’s
mind than any other device.

John Tukey [49]

328 Data visualization also means visualizing quantities and properties taken from
329 a series of events, in order to extract statistical meaning from them. Examples of
330 statistical data visualizations are histograms and scatter plots.

331 In HEP, like most other scientific disciplines, visualization of the data and of the
332 final results plays a key role in the analysis pipeline. A new projection of the data
333 may provide new insight, results must be summarized in a clear and concise way,
334 and multidimensional parameter spaces need to be visualized in an understandable
335 fashion. The discussion of how to properly display data is not new [50], but the tools
336 are constantly evolving. Since its introduction 20 years ago, ROOT [41] has become
337 the most widely used package in HEP to make plots, graphics, and even to build
338 event displays. It was developed at a time when there were few alternatives for the
339 HEP community that did not have a significant financial cost, and it has performed
340 admirably.

341 However, the landscape is changing and there are several existing tools driven
342 by non-HEP communities available. This section will look at some of the current
343 alternatives and comment on what options might be available in the future and
344 what our needs are. The main focus is on data exploration and presentation tools.
345 The former describes tools with which one prepares and builds visualizations for
346 the purpose of exploring and attempting to understand one’s dataset. The latter
347 describes tools for the presentation of final plots in a convenient and accessible way.
348 For more details on data analysis itself, one should refer to the HSF *Data Analysis*
349 and *Interpretation* Community White Paper [2].

350 **2.2.1 Desktop solutions**

351 As it stands, ROOT is the most widely adopted plotting tool within the HEP and
352 Nuclear Physics community. It has even made some inroads to the astrophysics
353 community and some small pockets within the financial community, to where some
354 physicists migrated. However, few other disciplines have adopted it. Still, it has
355 many features beyond the standard 1D/2D/3D histogram/graphing tools, such as
356 2D and 3D shapes, widgets for building a GUI, a JavaScript implementation for
357 web-based analysis [51] and is available within a Jupyter notebook [52]. But to
358 access the plotting features, an analyst must install the entire ROOT package which
359 includes file I/O, scientific libraries, fitting routines etc., and often the installation
360 process is non-trivial.

361 Many current HEP analysts make wide use of the Python programming language
362 and the PyROOT libraries [53]. Python is also very popular outside the HEP com-
363 munity and so it is worth looking at non-ROOT options available to Python users.
364 A recent (as of 2017) summary of the field was presented by Jake VanderPlas at
365 PyCon 2017 [54], a subset of which will be presented here. It is emphasized that this
366 is just a sampling and that the number of options available is a function of time.

- 367 • The Python library *matplotlib* [55], released in 2003, is the most mature plot-
368 ting tool for python and is the standard for most users. It can produce journal-
369 quality graphics and there are some add-ons that can improve the default plot-
370 ting options [56]. It does 1D, 2D, and 3D graphics with varying degrees of
371 success, but does not integrate with OpenGL libraries and so it can slow down
372 when the number of data points gets very large. It does produce most of the
373 histograms found in HEP but some minimal, extra work must be done by the
374 user to make histograms with error bars. Plots are reactive in the sense that
375 you can zoom in on different regions of the graph, but you cannot do anything
376 more significant with other mouseover commands (links, additional informa-
377 tion, etc.).
- 378 • The R programming language has several widely used graphics tools, both
379 built-in or provided by external modules. *ggplot2* [57] and *lattice* [58] are
380 particularly useful to visualize data in multidimensional parameter spaces. *gg-*
381 *plot2* is an implementation of the guidelines contained in the classic text *The*
382 *Grammar of Graphics* [59], while *lattice* is an implementation of the so-called
383 *Trellis Display* [60]. Both packages are very popular outside the HEP com-
384 munity and a wide range of learning materials are available in books, online
385 courses, and other media. Both packages are well developed and mature and
386 offer mechanisms for users to extend them by adding new features. *lattice* has
387 a longer history: in 2005, the year *ggplot2* first appeared, *lattice* was already

388 popular. In fact, figures made with *lattice* were shown in the presentation in
389 PHYSTAT05 [61] which introduced R to the particle physics community.

390 **2.2.2 Web-based solutions**

391 Web-based data visualization is also being rapidly developed. Very sophisticated
392 toolkits now provide tools to build web-based fully-responsive visualization of data
393 on all types of devices. In addition, they also offer other features, specially useful
394 for HEP, like full in-browser LaTeX rendering (with MathJAX) and real-time visu-
395 alization of streamed data. Being JavaScript-based, those libraries integrate with
396 the overall ecosystem of web-based technologies, letting them use all the tools of-
397 fered by other web libraries. They are overall a good solution for data presentation,
398 and can be combined with other tools such as Jupyter in order to be used for data
399 exploration. Some of the most used toolkits are described below:

- 400 • D3 (Data Driven Documents) [62] is perhaps the first web-based visualization
401 toolkit which has been widely adopted as the de-facto base solution for building
402 interactive data visualization for the web. The strong point of D3 is the link
403 of the data to the DOM entities and the possibility to work with SVG objects
404 natively. D3 is also the foundation layer upon which many higher level toolkits
405 are built.
- 406 • Bokeh [63]. This is a plotting utility from Continuum [64], the company behind
407 the *Anaconda* Python distribution system and other Python modules. It is
408 designed with the web in mind and builds in a high degree of interactivity into
409 the plots, making it useful to share results publicly and for building dashboards.
410 However, it works by writing HTML, which makes it difficult to work with
411 unless you use specific IDEs like a Jupyter notebook. Exporting a figure for a
412 journal article (*e.g.*, in the PNG format) is non-trivial as well, as that is not
413 currently the primary use-case for Bokeh.
- 414 • Plotly [65]. This is another web-oriented solution, similar to Bokeh and based
415 on the D3 library, where plots can be hosted in Plotly’s cloud service or viewed
416 in a Jupyter notebook. The plots are similarly very interactive and there are
417 ways to export figure images, but that is not the goal. Dashboards can be built
418 with relative ease and plotly offers libraries in R and JavaScript, in addition to
419 Python. They offer both a free and enterprise business model.

420 The so-called notebooks are a rapidly evolving way of using web-based technology
421 for both online and offline data analysis and visualization, with access to local re-
422 sources as well. After having started from a Mathematica-like notebook user interface
423 paradigm mixing server-side code snippet execution, structured text, and (mostly)
424 static visualizations, the Jupyter community is now exploring more interactive user

425 interface paradigms, including in the area of visualization. The JupyterLab project
426 is exploring a more MATLAB-like IDE user experience inside the web browser, with
427 features such as multiple source editing tabs and interactive python consoles. Its
428 ipywidgets sub-project tries to make Jupyter more interactive by moving more vi-
429 sualization work to client-side Javascript and introducing classic GUI widgets such
430 as sliders and checkboxes for interacting with the live visualization. The Belle II
431 experiment has started to use Jupyter notebooks to train new users in data analysis;
432 the learning curve is much gentler than in traditional terminal-based tutorials and
433 the time to useful visualization of results is much faster.

434 2.2.3 Issues

435 **Separating data visualization from the data analysis:** The suite of statistical
436 plotting tools in ROOT, Matplotlib, etc. are adequate for analysis, and their devel-
437 opment is very responsive to analysts' needs. However, it is often hard to separate
438 plot-making abilities from the data analysis framework. As a consequence, if a physi-
439 cist's data can only be found on a particular server, the plot-generating code must
440 also be located there and the outcome is sometimes hard to bring to the physicist's
441 laptop screen. In the worst cases, graphics files (PNGs) must be copied from the
442 server to the laptop for viewing. This causes a high interaction latency, discouraging
443 exploration. That's why the development of new tools should go towards a sharper
444 separation between the computation on data and the interactive data visualization
445 routines, pushing the latter to the client side as much as possible.

446 **Separating the plotting functions and content from the plotting style:**
447 Another symptom of the tight coupling between data analysis infrastructure and
448 plotting is that trivial changes to the plot— axis labels, colors, and such— are so
449 deeply buried in the analysis script that persistifying changes to them often requires
450 a full recalculation of the statistics. Changes to the final plot through the usage
451 of on-display user interfaces, in fact, are overwritten and lost if a plot is updated
452 for other reasons (new version of data upstream, for example). Here, one could use
453 inspiration from the increasing separation of logic and presentation that is occurring
454 in GUI toolkits (see *e.g.* use of CSS stylesheets in the GTK/GNOME environment).

455 A looser coupling between style and content, as well as a looser coupling be-
456 tween locality of computation and locality of rendering, would benefit the physics
457 community.

458 2.3 Non-spatial visualization

459 In HEP, there are data which are organized in a tree-like structure, and for which
460 a graph or a network visualization is the best choice. The Detector Description
461 is an example of a source of such data: it describes all the pieces which compose
462 a HEP experiment detector. The different pieces of the Detector Description are
463 interconnected through different relationships: geometrical volumes can be organized

464 in a parent-child relationship, or a property node can be shared among many volumes.
465 The visualization of those data in a network helps developers in the understanding
466 and the debugging of the Detector Description, by visualizing the relationships among
467 all the nodes and their properties. An example, from the ATLAS experiment [66],
468 of a graph visualizing the inner structure of a HEP detector description can be seen
469 in Figure 6.

470 Networks and graphs are very effective ways of visualizing tree-like data, because
471 they are able to show all the nodes, their relationships and their properties in a proper
472 way. Some degree of interactivity can let the scientists applying different filters and
473 layout, helping them to get rid of the clutter, to better understand and analyze the
474 data.

475 Another example of HEP data that can benefit from a graph-based visualization
476 is that one describing the execution chain of the jobs used to filter and reconstruct
477 the experimental data. Very recently HEP experiments began to develop new parallel
478 frameworks to concurrently handle analysis or reconstruction jobs, to efficiently
479 exploit the parallelism offered by the modern hardware (more details can be found
480 in the HSF *Event/Data Processing Frameworks* Community White Paper [67]). The
481 jobs are handled by a scheduler, which organizes them according to their needed
482 input and output data. The outcome of the scheduler is a directed acyclic graph
483 (DAG). The visualization by means of a graph helps the developer understanding
484 and debugging the reconstruction and framework code.

485 Other HEP experiments use graph-based tools to analyze and visualize other
486 types of data, like geographical distribution and load of computer networks used to
487 transfer data between GRID sites [68] (see Picture 7) or to store and query conditions
488 data [69].

489 All those data are not space- nor time-dependent, and they are better visualized
490 through a graph or a network. Graph-based visualization, as well as graph-databases,
491 are fairly new in the HEP landscape but they can be very powerful tools to effectively
492 visualize non-spatial data which are by their nature organized with a network layout.
493 We suggest the community further explores those tools, to better understand the
494 possibility offered by graph-based solutions for HEP needs.

495 3 Suggested guidelines and future development

496 As a community what we want to suggest here is the design and the usage of common
497 base visualization guidelines, to be able to share knowledge and best practices among
498 the different groups, and to foster collaboration among the HEP experiments.

499 3.1 A common community-defined format

500 Visualization has a key role in the lifecycle of a HEP experiment, addressing input
501 data from many sources and in many different formats. The input data are often

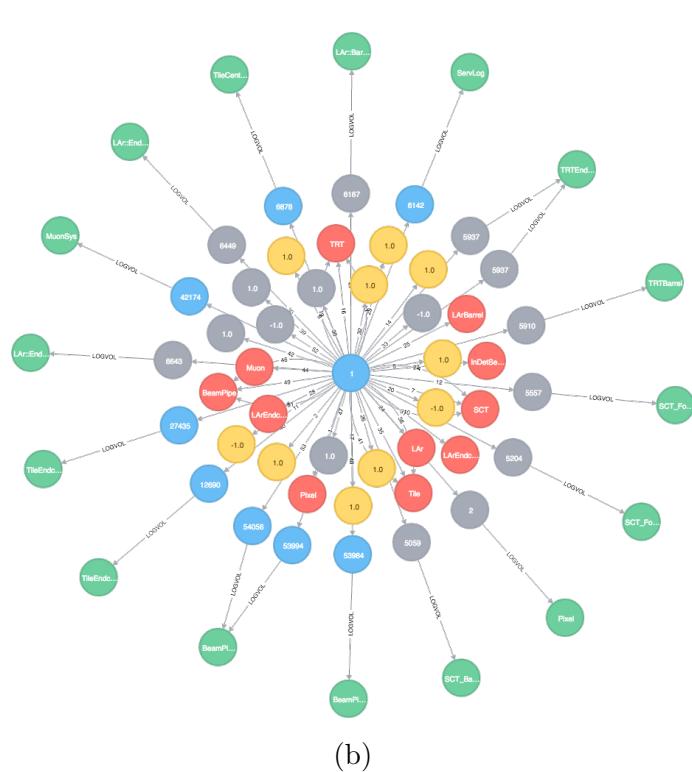
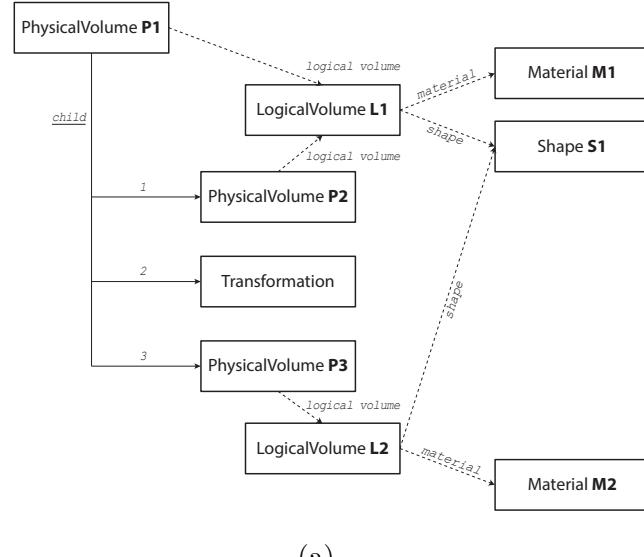


Figure 6: a) A schematic drawing depicting the tree-structure of the data describing the geometry of the ATLAS detector. Such data structures are better visualized using graphs and networks. b) A graph visualizing the first layer of the nodes of the ATLAS Detector Description. Different colors indicate different types of nodes; also, the labels along the lines state the different types of relationship between two data nodes [66].

502 quite tightly bound to a specific experiment's software framework and because of

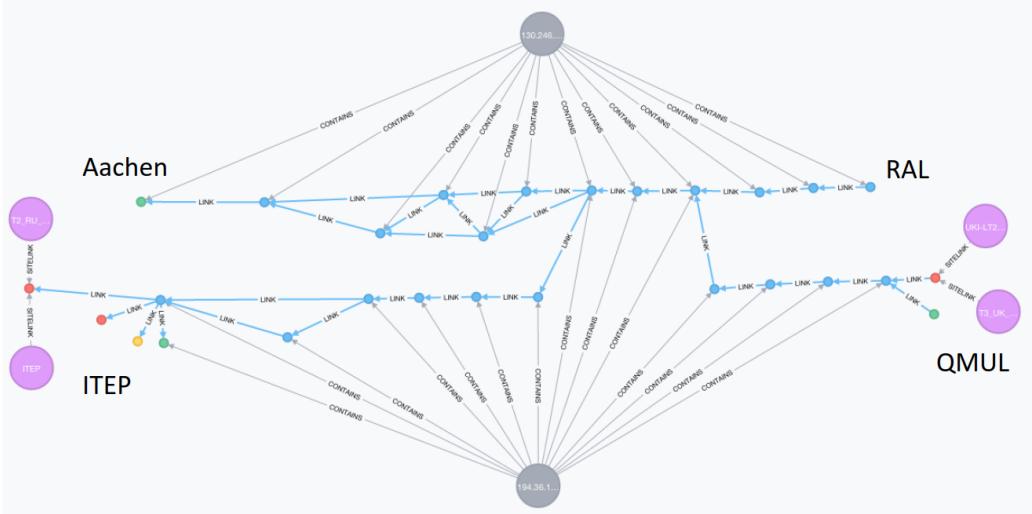


Figure 7: The image shows a graph used to visualize particular snapshot of the network topology between four WLCG sites, which was used to perform network path analysis. [Image provided by the authors of [68]].

that many visualization tools are integrated into the frameworks to some extent. However, the visualization is often the last step on the experiment data chain and the output of a visualization application is often not used by any other tool in the software framework. So while the direct interaction with experimental data formats is highly experiment specific, there is a real possibility of having the final stages of the visualization pipeline shared between several experiments.

Let us take the example of the detector geometry. There are very many different geometry libraries and formats in use among the HEP experiments. However, geometry libraries are all different ways to describe and handle basic geometrical entities and combinations of them. From a visualization point of view, the output of all geometry libraries are mere descriptions of 3D shapes, which could be abstracted from the underlying actual implementation. A shape like a box, or a cone, or a tube, or some boolean combination of them, could be interpreted and handled the same way by a visualization tool in all experiments.

The same reasoning made for the geometry can be done for the event data. Of course different experiments detect different objects and measure different quantities, but there are many common entities, especially among experiments within the same research field. For example, all experiments working on hadron colliders use the notion of particle track, which is usually constructed translating the track measurements into space points or points and angles, and use the notion of particle jet, usually visualized as a cone whose length is related to its energy and whose radius is linked to the specific algorithm used for the jet reconstruction. Both of these objects are currently often handled and visualized differently in different experiments,

526 but could in principle be the target of a common definition within the community.
527 If so, experiments could share best practices or snippets of code, if not complete
528 basic tools, to build and handle their visualization. In this way, the know-how and
529 the tools linked to visualization needs could be shared as well, and developed in a
530 collaborative way.

531 We as a community still need to propose and design such common definitions
532 and guidelines. This will be addressed in the second phase of this Community action,
533 following the completion of the Community White Paper.

534 For the moment, we observe that several event displays in experiments have
535 exporters to translate geometry information to standard formats used in the com-
536 munities outside HEP, mainly in computer graphics and engineering. For example
537 the Belle II experiment has written exporters from Geant4 data to different formats,
538 including VRML [70] and FBX [71], which are two of the most common formats used
539 to store and share 3D graphics data. The Unity game-development engine [37], in
540 turn, can export the FBX geometry to the glTF format [72], an emerging royalty-free
541 specification for 3D objects and scenes, for fast web distribution via the SketchFab
542 community repository [73, 74]. Displays based on three.js have access to multiple
543 importers and exporters of several geometry formats, including the ones mentioned
544 above.

545 In order to start sharing knowledge and to start working on demonstrators to
546 show and share best practices, we propose to start defining a common format to ex-
547 change data among the experiments. We should start by finding and listing common
548 shared objects from geometry and event data. After that, we should start converg-
549 ing on a shared definition of those objects, to build a common design toward a data
550 model to handle and serve them. The idea, in fact, is to enable usage of this common
551 format to visualize data from the different experiments with the same shared best
552 practices, if not the same foundation software tools.

553 We think that community-developed common formats and tools should also be
554 extendable, to let the experiments add their own custom content and objects. As
555 an example, calorimeter cells can be of very different shapes, and an experiment
556 might need to add its own custom shapes to the common format to visualize them
557 properly. Thus, in addition to the part handling the common objects, there should
558 be a part of the format targeted at storing extended custom content, specific to
559 a given experiment. For such custom content, experiments will have to develop
560 custom visualization tools as well; but they could build them upon the foundation
561 of the community-driven part.

562 Some experiments in that direction have been performed within the community
563 in the already. For example, the ALICE experiment made use of the mini “Visual-
564 ization Summary Data” (VSD) set of classes, contained in the ROOT Event Visual-
565 ization Environment (EVE), to make ALICE data visualization decoupled from the
566 AliROOT experiment’s framework [75].

567 **3.2 Serving the geometry and event data through services**

568 Once a common format for shared objects is defined, we believe that the design and
569 the development of online services to query and serve the geometry data would be
570 a very useful addition to the landscape. The main driving force is the realization
571 that detector description should be much more accessible than it is today. For many
572 experiments, accessing the detector description means starting and running at least
573 parts of the experiment’s framework. The need of accessing a specific geometry
574 version, in fact, is critical for reconstruction and simulation. However, the geometry
575 data needed for event visualization can often be simpler: even when showing the
576 actual geometry of the experiment, accessing the latest alignment constants is not
577 crucial for visualization purposes, because small differences in the geometry are rarely
578 visible in an event display. So, we think that serving a “frozen” version of the
579 experiment’s geometry would be enough, and that a simpler way to retrieve it should
580 be designed, to ease data access for visualization: for example, an online service could
581 remotely serve the geometry data to the visualization applications in a standardized
582 format.

583 It would be desirable for the new mechanism to have a search/filter functionality
584 too, to let client applications query for a specific subset of information and for a
585 way to select the level of details, to set the desired accuracy and complexity of the
586 retrieved geometry.

587 The same could be envisaged for event data, even though that is a more com-
588 plicated task, involving many different layers and services: very often event data are
589 stored on the Grid and very often they need to be processed in order to be usable
590 for event displays. However, experimental data should be made more accessible for
591 visualization. Thus, in collaboration with the HSF *Data Access and Management*
592 working group, an API or a service to get streamed event data will be designed. In
593 addition, simulation data description for visualization could be handled in the same
594 way, by implementing converters from generators and simulation applications.

595 After a first phase of development and stand-alone testing, the streamed data
596 could be used by the current visualization tools as well, as the first step of their
597 modernization and towards the usage of common community-developed techniques
598 and best practices.

599 **3.3 Client-server architecture for geometry and event data visualization**

600 After common data formats and mechanisms to serve them are designed, together
601 with a set of exporters required to translate the experiments’ data to the common
602 format, we are proposing to build a client-server architecture, upon which next gen-
603 eration visualization applications can be built.

604 The idea behind that is that if we can send commands from the client to the
605 server, and get the answer back in the data stream, then we will be able to interact

with the experiment’s framework as well, in addition to using common visualization applications to visualize the common objects. In this way, we can develop a modular architecture where HEP experiments can share the design, the development and the maintenance of common visualization tools, while maintaining a certain degree of freedom to add custom content and objects and to interact with their own framework to retrieve specific content.

3.4 Exploring modern technologies

3.4.1 Graphics and game engines

So far direct OpenGL, its derivatives (like WebGL), or old graphics libraries have been used in HEP visualization applications. Nowadays, another type of graphics library is rapidly evolving, those embedded in so-called game engines. Game engines are software frameworks targeted at the gaming industry, and they feature very efficient, optimized, and modern 3D graphics. Integration with existing code is not easy, however, because they are usually meant to be used as development environments, and not as embedded libraries like those commonly used in HEP. So they would probably require some major changes in the usual software architecture used in HEP. But they offer very optimized graphics and modern features, like tools for Virtual Reality, which could be exploited in our applications.

Some HEP experiments have recently started to successfully use them to build visualization applications and event displays, like Belle II [39] and the Total Event Visualiser (TEV) of the CERN Media Lab [76], which used the Unity game engine [37], and ATLAS which used the Unreal Engine [77] for its virtual reality application ATLASrift [38]. These two game engines are the most popular ones on the market, and they are free for educational and non-commercial projects.

Unreal Engine is fully open source and it supports two modes of development (C++ and the so-called Blueprints [78]) that can be used interchangeably even in the same project. It produces extremely performant executables for basically all platforms (Windows, OSX, Linux, iOS, Android, Web, all VR platforms). All parts of development cycle are fast even for a novice, thanks to powerful tools implemented as plugins. They have large developer communities and are very fast in supporting the latest technologies; for example it already supports Vulkan [79], the cross-platform 3D and computing API.

The Unity development platform [37] is very intuitive for novices as well as experts and provides rapid turnaround during the development cycle: the project can be executed immediately without having to compile and link an executable. All of the platforms supported by Unreal are supported as targets by Unity as well. Presently, user code is written in either C# or an adaptation of Javascript.

Another game engine that has gained attention in recent years is the open-source engine Godot [80]. While it is still not quite at the same level as Unity or Unreal

645 Engine, it allows the deployment to similar platforms as Unity and Unreal Engine
646 but is very lightweight. It offers support for 2D and 3D graphics and for multiple
647 programming languages *e.g.* GDScript (a Python-like scripting language), C# 7.0 (by
648 using Mono), and C++. It also offers visual scripting using blocks and connections
649 and support for additional languages with community-provided support for Python,
650 Nim [81], D and other languages.

651 As a community, we would like to explore further the features those modern
652 game engines can offer. Also, we would like to take a look at possible usage patterns
653 in the context and within the workflow of HEP visualization.

654 Finally, another new entry in the 3D graphics engines landscape is Qt3D [82].
655 The key feature of Qt3D is that it is natively integrated with the Qt framework,
656 which eliminates a layer which was needed until now, *i.e.*, a glue package to con-
657 nect the Qt GUI with the window showing the 3D content (like, for example, the
658 SoQt [83] package used by ATLAS). By eliminating that, we could simplify the ar-
659 chitecture of our visualization tools and lower the maintenance workload. Qt3D is
660 still in development, but it shows an initial set of features which are worth a further
661 consideration of the new toolkit. We plan to take a look at its development in the
662 near future, to see if it can satisfy the HEP requirements. Also, being open source,
663 we could consider contributing to the Qt3D software project as a community, by
664 providing the pieces we need for our applications.

665 **3.4.2 Web-based applications**

666 Web-based graphics have traditionally been considered not powerful enough to handle
667 the thousands of volumes that can be shown in HEP event displays, for example,
668 when visualizing hits in a very busy event. But the technology has rapidly evolved
669 and web-based graphics can now visualize very complex scenes. For example, the
670 glTF [72] 3D model of the Belle II detector [84], with tens of thousands of elements,
671 can be loaded, viewed and manipulated in a web browser, even on a smartphone,
672 very effectively [73].

673 The advantages of visualization in the browser have been mentioned previously
674 and several application have already been developed. There is therefore already a
675 strong interest in the community in supporting the usage and the development of
676 web-based tools. In particular, JSROOT [51] could be used as an underlying layer
677 for event data visualization as well as three.js [34], which have been used successfully
678 by different experiments (*e.g.*, [28, 30, 31]) to visualize geometry and event data.

679 **3.4.3 Virtual and augmented reality**

680 As briefly described in Section 2.1.2, Virtual Reality (VR) describes the simulation
681 of the user’s physical presence in a virtual environment. This simulation is typi-
682 cally delivered via a Head Mounted Display (HMD) that provides visual and aural
683 experience of the simulated world. Rotational and positional tracking of the user’s

684 head and hand motion (when available) allow for interaction and navigation in the
685 virtual environment. This lets the user live an immersive experience of the virtual
686 world offering many degrees of freedom. Purely rotational motion is usually referred
687 to as 3 degree-of-freedom motion; when combined with positional motion support,
688 an application is said to support 6 degrees-of-freedom.

689 There are several ways to deliver VR to the user with varying levels of functional-
690 ity, accessibility, and cost. They range from applications running on a mobile phone
691 viewed through simple headsets to the most realistic and immersive VR experiences
692 provided by the combination of advanced HMDs and desktop computers.

693 The simplest and most inexpensive way to deliver VR is via the web browser
694 on a mobile phone and viewed through a Google Cardboard [85] headset (which can
695 itself be literally made from cardboard). Rotational tracking is achieved through
696 device orientation controls, either in a native application or using the HTML5 device
697 orientation API in the browser. No hand controller is used in the Cardboard but for
698 native applications a click event is available via a magnet attached to the Cardboard
699 viewer.

700 The Google Daydream [86] is the next iteration of viewers developed by Google
701 for their Google VR technology. Content is still delivered by a mobile phone (thus,
702 the performance is limited by the computing power of the phone) but a Bluetooth-
703 connected hand controller with rotational tracking is available. The Samsung Gear
704 VR platform [87] is another headset powered by an inserted smartphone targeted for
705 the Oculus platform.

706 Currently, the most immersive and interactive VR environments are provided by
707 the Oculus Rift [88] and the HTC Vive [89], which combine the computing resources
708 of a desktop machine with sensors, controllers, and high-quality HMDs. Those so-
709 phisticated devices are relatively expensive (even if prices are lowering in the last
710 months) and require a fairly powerful computer to run. The presence of a proper
711 computer solves the performance issue of phone-based viewers, since the 3D graphics
712 computations are handled by the computer, but that also increases the required ini-
713 tial budget for people willing to test such technologies, and that limits the number
714 of people getting access to such platforms, for example in public events organized by
715 HEP institutes.

716 Meanwhile, a new type of device has been recently developed: a standalone
717 viewer, equipped with an on-board CPU, able to run medium computation-intensive
718 applications. Those devices lower the budget needed to develop and deploy VR
719 applications significantly. Examples of those new devices are Oculus Go[90] and the
720 stand-alone Lenovo headset for the Daydream environment [91].

721 Game engines, described in Section 3.4.1, provide powerful integrated devel-
722 opment environments for creation of VR applications targeting multiple devices.
723 Thanks to engines' abstraction of third party VR libraries, most HEP experiments
724 should be able to develop VR applications which natively support both standard

725 displays and all VR hardware. Currently, both ATLASrift [38] and Belle II VR [39]
726 support Oculus, HTC Vive, and standard 2D displays. CMS, using the Unity game
727 engine, is currently working on CMS.VR (to be released), targeting Oculus and HTC.

728 Augmented Reality (AR) applications use the device’s camera to looks at the
729 world around the user, to use that as an underlying layer, over which, based on the
730 user’s interaction and motion, they dynamically render simulated virtual entities.
731 The user can navigate in the real world, while looking at and interacting with the
732 virtual objects. AR technology can be used in HEP as an educational tool, for
733 instance to dynamically show and describe a HEP detector to a group of people or
734 a class.

735 ATLAS and ALICE researchers have started to explore the possibilities offered
736 by augmented reality for outreach and education, by using Unity [37] and the Vu-
737 foria framework [92] (commercial, but free for development). The ATLAS-in-Your-
738 Pocket [93] application uses printed marks to place a rendered geometry of the AT-
739 LAS detector on top of a view of the real world in front of the user. The More-Than-
740 ALICE [94] application allows users to superimpose a description of the detector or
741 event visualizations to the camera image of the actual ALICE detector (for example,
742 on a screen or during public visits to the experimental site) or its paper model.

743 Development of web-based VR and AR applications for mobile browser can be
744 done using a WebGL library such as three.js [34] and using the HTML device orien-
745 tation control API. The viewport is split into two views, one for each eye, each with a
746 dedicated camera view separated by an appropriate distance to create a stereoscopic
747 effect (*e.g.* iSpy WebGL [28] has a stereo mode for Google Cardboard). The develop-
748 ing WebVR specification [95] provides interfaces to VR hardware via the browser. A
749 powerful framework for development of VR applications for various devices using the
750 browser is A-Frame [96]. A-Frame has support for several device controllers as well,
751 such as for the Daydream and Oculus. CMS is exploring the possibilites of A-Frame
752 in the browser for both VR and AR (*e.g.* CMS A-Frame prototype [97]).

753 VR and AR applications could be used, in principle, to build event displays
754 and data visualization tools for research work as well. However, the current data
755 access model prevents a straightforward use of those technologies together with the
756 experiments’ software frameworks. A future client-server approach to data access
757 could fill the current gap and let developers build new VR and AR tools targeting
758 HEP physicists as end users.

759 **3.4.4 Mobile technologies**

760 Portability and simplicity of usage are the strong points of mobile devices. More
761 than as “mobile” devices, smartphones, tablets and ultrabooks can be considered,
762 as devices “close to people”. As such, the usage of such devices should be exploited
763 more in the final steps of the visualization chain, where heavy batch data processing
764 is not needed. For instance, their usage should be leveraged for the production

765 and visualization of event displays. Ideally, a user should be able to easily retrieve
766 interesting events from the experiment and interactively visualize them on all kinds
767 of devices.

768 That is why we strongly promote the usage of the server-client architecture
769 described and supported in this paper in Section 3.3 and the new data access patterns
770 presented and supported in the the HSF *Data Access and Management* Community
771 White Paper [98]. This would open up new possibilities for interactive visualization
772 on mobile devices: it would let visualization clients running on mobile devices connect
773 to server tools running in the experiment’s framework to easily and interactively
774 retrieve the desired data.

775 It is worth noting that in other areas of science, for instance in astronomy,
776 researchers have worked to facilitate data access and to migrate to more standard
777 data formats. This allowed for the possibility of having data visualization tools on
778 mobile devices, in addition to desktop and laptop machines. Moreover, this not only
779 helped the researchers in accessing and visualizing their data, but it also helped in
780 making science accessible by the public, having eased the development of programs
781 used in Outreach and Education activities and events. It is true that HEP data are
782 usually much more complex than astronomy data, and so it will be harder to achieve,
783 but we think that an effort in simplifying the access to experimental data would be
784 worthwhile.

785 Therefore, the leverage of the usage of mobile devices in HEP adds a strong
786 point to the development and the support of common client-server tools and data
787 exchange formats among HEP experiments in the near future.

788 **3.4.5 Multi-user applications**

789 Nowadays multi-user technology is used in many applications: for example in Google-
790 Docs, where many users can simultaneously interact with the same document. What
791 we would like to provide is multi-user support for visualization to let several users
792 explore and interact with events at the same time. Beside being a useful feature for
793 expert users (for example, when asking for advice on the visualization of a piece of
794 detector to another person at a distant institute), it could be important for outreach
795 and education activities, where students and other people could interact together
796 with an event display, or for virtual guided tours. Game engines offer multi-user
797 support natively, thus we could start exploring their usage. An example of such
798 collaborative features in a 3D environment is integrated in the Med3D visualization
799 framework [99].

800 **4 Sharing knowledge and fostering collaboration**

801 During the kick-starter meetings and the different workshops organized to start and
802 develop the present Community White Paper, the whole HSF Visualization Working

803 Group has agreed on the importance of sharing knowledge among the whole HEP
804 community, as well as best practices and know-how. Too often, in fact, solutions
805 and tools developed for one HEP experiment are not sufficiently advertised to the
806 rest of the HEP Visualization community, with the result that the community base
807 knowledge is fragmented and not efficiently exploited.

808 The focus of this Working Group in fact is not limited to the preparation of this
809 white paper. Instead, a longer term collaboration among the experiments is foreseen,
810 in order to collaborate on common visualization projects. To foster collaboration and
811 sharing, we agree on following courses-of-action which are described below.

812 **4.1 Yearly workshop**

813 On March 2017 the first HSF Visualization Workshop was organized at CERN to
814 let all the experts from the different experiments and projects show their work and
815 share their solutions. In addition, external experts from industry were invited to
816 present the latest advancements in the field and best practices [100].

817 It was the first topical workshop focused on HEP Visualization in many years.
818 Many HEP experiments and communities showed their latest developments. Given
819 the high number of presentations, a second mini-workshop [101] has been organized
820 as a follow-up, to let the remaining communities to present their work.

821 The Working Group agreed on the importance of meeting to share findings,
822 knowledge and solutions; and it was decided to try to organize a topical workshop
823 on HEP visualization once per year.

824 An important point was raised while organizing the first Workshop: other sci-
825 entific fields have visualization and graphics needs similar to HEP, for example Geo-
826 physics. In future workshops we will try to have presentations from other communi-
827 ties as well, in order to try to foster friendly and fruitful collaborations which could
828 benefit the whole scientific community.

829 **4.2 Code repository**

830 The HSF Visualization Working Group also agreed on the importance of fostering
831 collaborative work. As a start a new dedicated project has been created within the
832 HSF GitHub repository [102]. It is intended to be the space where members of the
833 Working Group can share their work-in-progress studies and their solutions, and
834 where community-driven projects will be stored.

835 **5 Roadmap**

836 **5.1 One year**

837 In the first year the Visualization Working Group will work on defining R&D projects,
838 based on the key points and ideas discussed in this community white paper.

839 The main goal will be developing techniques and tools which let visualization
840 applications and event displays be less dependent on specific experiments' software
841 frameworks, leveraging the usage of common packages and common data formats.

842 In a first phase, the community will identify common objects and will agree on
843 common definitions, as described in section 3.1. Then, a common data exchange
844 format, either based on custom data formats or, if possible, on open standards, will
845 be designed by the community. After that, exporters and interface packages would
846 be designed as bridges from the experiments' frameworks, which are needed to access
847 data at a high level of detail, to the common packages.

848 **5.2 Three years: ATLAS and CMS Computing TDRs**

849 In the second and third year the Visualization WG will work on designing and build-
850 ing demonstrators to show the feasibility of the community-driven best practices and
851 tools. The goal will be to get a final design of those tools, to be included in the de-
852 velopment plans of the different experiments. Moreover, the WG will work towards
853 a more convenient access to geometry and event data. In collaboration with the HSF
854 *Data Access and Management* working group, an API or a service to get streamed
855 event data would be designed.

856 **5.3 Five years: Towards HL-LHC**

857 In the fourth and fifth year, the focus will be on developing the actual community-
858 driven tools, to be used by the experiments for their visualization needs in production.

859 The goal will be the usage of the community-developed tools within the experi-
860 ments' visualization applications; and perhaps the usage of a simplified data access,
861 but that depends on the actual feasibility, which will be established after an initial
862 study.

863 **6 Conclusions**

864 Modern and modular visualization tools, which will feature simplified data access
865 and retrieval as well, would leverage the accessibility, letting end users exploit all the
866 possibilities offered by modern visualization solutions, without the need of running
867 them on specific platforms, running them within the experiments' software frame-
868 works, or being bound to specific solutions. And a better experience will reflect to
869 a better usage, which will positively affect the usage of such visualization tools for
870 detector development and simulation of new experiments, as well as the data analysis
871 and the upgrade studies of the current ones.

872 In the end, common community-driven tools will let users of all experiments make
873 use of the latest and best tools, while sharing the development, the maintenance and
874 the workload among all the experiments.

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