Addressing Big Data Challenges for Scientific Data Infrastructure

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Abstract—This paper discusses the challenges that are imposed by Big Data Science on the modern and future Scientific Data Infrastructure (SDI). The paper refers to different scientific communities to define requirements on data management, access control and security. The paper introduces the Scientific Data Lifecycle Management (SDLM) model that includes all the major stages and reflects specifics in data management in modern e-Science. The paper proposes the SDI generic architecture model that provides a basis for building interoperable data or project centric SDI using modern technologies and best practices. The paper explains how the proposed models SDLM and SDI can be naturally implemented using modern cloud based infrastructure services provisioning model.

Keywords- Big Data Science, Scientific Data Infrastructure (SDI), Scientific Data Lifecycle Management (SDLM), Cloud Infrastructure Services.

I. INTRODUCTION

Modern e-Science infrastructure allows targeting new large scale problems which solution was not possible before, e.g. genome, climate, global warming, etc. e-Science typically produces a huge amount of data that need to be supported by a new type of e-Infrastructure capable to store, distribute, process, preserve, and curate these data: we refer to this new infrastructures as Scientific Data e-Infrastructure (SDI) [1, 2].

In e-Science, the scientific data are complex multifaceted objects with the complex internal relations, they are becoming an infrastructure of their own and need to be supported by corresponding physical or logical infrastructures to store, access and manage these data.

The emerging SDI should allow different groups of researchers to work on the same data sets, build their own (virtual) research and collaborative environments, safely store intermediate results, and later share the discovered results. New data provenance, security and access control mechanisms and tools will allow researchers to link their scientific results with the initial data (sets) and intermediate data to allow future re-use/re-purpose of data, e.g. with the improved research technique and tools.

This paper analyses the new challenges imposed to modern e-Science infrastructures by the emerging big data technologies; it proposes a general approach and architecture solutions that introduce the new Scientific Data Lifecycle Management (SDLM) model and the generic SDI architecture model that provides a basis for heterogeneous SDI components interoperability and integration, in particular based on cloud infrastructure technologies.

The paper is organised as follows. Section 2 gives and overview of the main research communities and summarises requirements to future SDI. Section 3 discusses challenges to data management in Big Data Science, including SDLM discussion. Section 4 introduces the proposed e-SDI architecture model that is intended to answer the future big data challenges and requirements. Section 5 discusses SDI implementation using cloud technologies. Section 6 discusses specific requirements and provides suggestions about building federated access control infrastructure for modern and future SDI.

II. GENERAL REQUIREMENTS TO BIG DATA E-SCIENCE INFRASTRUCTURE

A. Paradigm change in Big Data Science

Big Data Science is becoming a new technology driver and requires re-thinking a number of infrastructure components, solutions and processes to address the following general challenges [2, 3]:

- Exponential growth of data volume produced by different research instruments and/or collected from sensors
- Need to consolidate e-Infrastructure as persistent research platform to ensure research continuity and cross-disciplinary collaboration, deliver/offer persistent services, with adequate governance model.

The recent advancements in the general ICT and big data technologies facilitate the paradigm change in modern e-Science that is characterized by the following features:

- Automation of all e-Science processes including data collection, storing, classification, indexing and other components of the general data curation and provenance
- Transformation all processes, events and products into digital form by means of multi-dimensional multi-faceted measurements, monitoring and control; digitising existing artifacts and other content.
- Possibility to re-use the initial and published research data with possible data re-purposing for secondary research
- Global data availability and access over network for cooperative group of researchers, including wide public access to scientific data.
• Existence of necessary infrastructure components and management tools that allows fast infrastructures and services composition, adaptation and provisioning on demand for specific research projects and tasks.

• Advanced security and access control technologies that ensure secure operation of the complex research infrastructures and scientific instruments and allow creating trusted secure environment for cooperating groups and individual researchers.

The future SDI should support the whole data lifecycle and explore the benefit of the data storage/preservation, aggregation and provenance in a large scale and during long/unlimited period of time. Important is that this infrastructure must ensure data security (integrity, confidentiality, availability, and accountability), and data ownership protection. With current needs to process big data that require powerful computation, there should be a possibility to enforce data/dataset policy that they can be processed on trusted systems and/or complying other requirements. Researchers must trust the SDI to process their data on SDI facilities and be ensured that their stored research data are protected from non-authorised access. Privacy issues are also arising from distributed remote character of SDI that can span multiple countries with different local policies. This should be provided by the Access Control and Accounting Infrastructure (ACAI) which is an important component of SDI [4, 5].

B. Research communities and SDI requirements

In this section we provide short overview of the research infrastructures and communities, in particular defined for the Europe Research Area (ERA) [3] and analyse their specific requirement to future SDI to address Big Data challenges.

We refer to existing studies of European e-Infrastructures that analyse the scientific communities practices and requirements such as those undertaken by the SIENA Project [6], EIROforum Federated Identity Management Workshop [5], European Grid Infrastructure (EGI) Strategy Report [7], UK Future Internet Strategy Group Report [8].

The High Energy Physics community represents a large number of researchers, unique expensive instruments, huge amount of data that are generated and need to be processed continuously. This community has already operational Worldwide Large Hadron Collider Grid (WLCG) [9] infrastructure to manage and access data, protect their integrity and support the whole scientific data lifecycle. WLCG development was an important step in the evolution of European e-Infrastructure that currently serves multiple scientific communities in Europe and internationally. The EGI cooperation [7] manages European and worldwide infrastructure for HEP and other communities.

Material science, analytical and low energy physics (proton, neutron, laser facilities) is characterised by short projects, experiments and consequently highly dynamic user community. It requires highly dynamic supporting infrastructure and advanced data management infrastructure to allow wide data access and distributed processing.

Environmental and Earth science community and projects target regional/national and global problems. They collect huge amount of data from land, sea, air and space and require ever increasing amount of storage and computing power. This SDI requires reliable fine-grained access control to huge data sets, enforcement of regional issues, policy based data filtering (data may contain national security related information), while tracking data use and keeping data integrity.

Biological and Medical Science (also defined as Life science) have a general focus on health, drug development, new species identification, new instruments development; they generates massive amount of data and new demand for computing power, storage capacity, and network performance for distributed processes, data sharing and collaboration. Biomedical data (healthcare, clinical case data) are privacy sensitive data and must be handled according to the European policy on Personal Data processing [10].

Social Science and Humanities community and projects are characterised by multi-lateral and often global collaboration between researchers from all over the world who need to be engaged into collaborative groups/communities and supported by collaborative infrastructure to share data, discovery/research results and cooperatively evaluate results. The current trend to digitise all currently collected physical artifacts will create in the near future a huge amount of data that must be widely and openly accessible.

The following are general infrastructure requirements to SDI for emerging Big Data Science:

• Support long running experiments and large data volumes generated at high speed
• Multi-tier data distribution and replication
• Support of virtual scientists communities
• Trusted environment for data storage and processing
• Data integrity, confidentiality, accountability
• Policy binding to data to protect privacy

III. DATA MANAGEMENT IN BIG DATA SCIENCE

Emergence of computer aided research methods is transforming the way how research are done and scientific data are used. The following types of scientific data are defined [4]:

• Raw data collected from observation and from experiment (according to an initial research model)
• Structured data and datasets that went through data filtering and processing (supporting some particular formal model)
• Published data that supports one or another scientific hypothesis, research result or statement
• Data linked to publications to support the wide research consolidation, integration, and openness.

Once the data is published, it is essential to allow the other scientists to be able to validate and reproduce the data that they are interested in, and possibly contribute with new results. Capturing information about the processes involved in transformation from raw data up until the generation of published data becomes an important aspect of scientific data management. Scientific data provenance becomes an issue
that also needs to be taken into consideration by SDI providers [11].

Another aspect to take into consideration is to guarantee reusability of published data within the scientific community. Understanding semantic of the published data becomes an important issue to allow for reusability, and this has been traditionally been done manually. However, as we anticipate unprecedented scale of published data that will be generated in Big Data Science, attaching clear data semantic becomes a necessary condition for efficient reuse of published data. Learning from best practices in semantic web community on how to provide a reusable published data, will be one of consideration that will be addressed by SDI.

Big data are typically distributed both on the collection side and on the processing/access side: data need to be collected (sometimes in a time sensitive way or with other environmental attributes), distributed and/or replicated. Linking distributed data is one of the problems to be addressed by SDI.

The European Commission’s initiative to support Open Access to scientific data from publicly funded projects suggests introduction of the following mechanisms to allow linking publications and data [12, 13]:

- PID - persistent data ID
- ORCID – Open Researcher and Contributor Identifier [14].

The required new approach to data management and handling in e-Science is reflected in the Scientific Data Lifecycle Management (SDLM) model (see Figure 1) we propose based on the result of analysis of the existing practices in different scientific communities. The proposed model is compliant with the data lifecycle study results presented in [15].

The generic scientific data lifecycle includes a number of consequent stages: research project or experiment planning; data collection; data processing; publishing research results; discussion, feedback; archiving (or discarding).

The dynamic character of SDI and its support of distributed multi-faceted communities are supported by the dedicated layers: D3 – Infrastructure Virtualisation layer that typically uses modern cloud technologies; and D5 – Federation and policy layer that incorporates related federated infrastructure management and access technologies [4, 16, 17]. Introducing the Federation and Policy layer reflects current practice in building and managing complex SDI (and also enterprise infrastructures) and allows independently managed infrastructures to share resources and support the inter-organisational cooperation.

IV. PROPOSED SDI ARCHITECTURE MODEL

We also propose SDI Architecture for e-Science (e-SDI) is illustrated in Figure 2. It contains the following layers:

Layer D1: Network infrastructure layer represented by the general purpose Internet infrastructure and dedicated network infrastructure

Layer D2: Datacenters and computing resources/facilities

Layer D3: Infrastructure virtualisation layer that is represented by the Cloud/Grid infrastructure services and middleware supporting specialised scientific platforms deployment and operation

Layer D4: (Shared) Scientific platforms and instruments specific for different research areas

Layer D5: Federation and Policy layer that includes federation infrastructure components, including policy and collaborative user groups support functionality.

Layer D6: Scientific applications and user portals/clients
VI. CLOUD BASED INFRASTRUCTURE SERVICES FOR SDI

Figure 3 illustrates the typical e-Science or enterprise collaborative infrastructure that is created on demand and includes enterprise proprietary and cloud based computing and storage resources, instruments, control and monitoring system, visualization system, and users represented by user clients and typically residing in real or virtual campuses.

The main goal of the enterprise or scientific infrastructure is to support the enterprise or scientific workflow and operational procedures related to processes monitoring and data processing. Cloud technologies simplify building of such infrastructure and provision it on-demand. Figure 3 illustrates how an example enterprise or scientific workflow can be mapped to cloud based services and after that deployed and operated as an instant inter-cloud infrastructure. It contains cloud infrastructure segments IaaS (VR3-VR5) and PaaS (VR6, VR7), separate virtualised resources or services (VR1, VR2), two interacting campuses A and B, and interconnecting them network infrastructure that in many cases may need to use dedicated network links for guaranteed performance.

Figure 3. From scientific workflow to cloud based infrastructure.

Efficient operation of such infrastructure will require both overall infrastructure management and individual services and infrastructure segments to interact between themselves. This task is typically out of scope of the existing cloud service provider models but will be required to support perceived benefits of the future e-SDI. These topics are a subject of another research by authors on defining the InterCloud Architecture Framework [18, 19].

VI. FUTURE RESEARCH AND DEVELOPMENT

The future research and development will include further SDLM definition, e-SDI and ACAI components definition and development with focus on infrastructure components of e-SDI. Special attention will be given to defining the whole cycle of the provisioning SDI services on-demand specifically tailored to support instant scientific workflows using cloud IaaS and PaaS platforms. This research will be also supported by development of the corresponding Cloud and InterCloud architecture framework to support Big Data e-Science processes and infrastructure operation.

REFERENCES


