

1 Networks in ATLAS

2 For the ATLAS Collaboration, Shawn McKee¹

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6 **Abstract.** Networks have played a critical role in high-energy physics (HEP), enabling us
7 to access and effectively utilize globally distributed resources to meet the needs of our
8 physicists.

9 Because of their importance in enabling our grid computing infrastructure many physicists
10 have taken leading roles in research and education (R&E) networking, participating in, and
11 even convening, network related meetings and research programs with the broader networking
12 community worldwide. This has led to HEP benefiting from excellent global networking
13 capabilities for little to no direct cost. However, as other science domains ramp-up their need
14 for similar networking it becomes less clear that this situation will continue unchanged.

15 What this means for ATLAS in particular needs to be understood. ATLAS has evolved its
16 computing model since the LHC started based upon its experience with using globally
17 distributed resources. The most significant theme of those changes has been increased reliance
18 upon, and use of, its networks.

19 We will report on a number of networking initiatives in ATLAS including participation in
20 the global *perfSONAR* network monitoring and measuring efforts of WLCG and OSG, the
21 collaboration with the LHCOPN/LHCONE effort, the integration of network awareness into
22 PANDA, the use of the evolving ATLAS analytics framework to better understand our
23 networks and the changes in our DDM system to allow remote access to data.

24 We will also discuss new efforts underway that are exploring the inclusion and use of
25 software defined networks (SDN) and how ATLAS might benefit from:

- 26 • Orchestration and optimization of distributed data access and data movement.
- 27 • Better control of workflows, end to end.
- 28 • Enabling prioritization of time-critical vs normal tasks
- 29 • Improvements in the efficiency of resource usage

30 1. Introduction

31 Innovation supporting science continues to increase requirements for the computing and networking
32 infrastructures of the world. Instrumentation, storage, processing facilities and collaborative partners
33 are often geographically and topologically separated, thus complicating the problems involved with
34 data management. Global scientific collaborations, such as ATLAS , continue to push the network
35 requirements envelope. Data movement in this collaboration is routinely including the regular
36 exchange of many 10's of petabytes of datasets between the collection and analysis facilities in the
37 coming years. This increased emphasis on the “network”, now a vital resource on par with the actual
38 scientific process, implies that it **must** be a highly capable and reliable resource to ensure success; the

39 lack thereof could mean critical delays in the overall scientific progress of distributed data-intensive
40 experiments.

41 We will report on the role of networking in supporting the scientific mission and goals of the
42 ATLAS collaboration. Networks are fundamental to the distributed computing model ATLAS has
43 developed and, as such, end-to-end network performance and network problems have a significant
44 impact on the ability of ATLAS physicists to reach their scientific goals in a timely manner. In this
45 paper we will discuss the ongoing efforts to monitor, measure and maintain our networks and
46 exploratory work to integrate programmable networks into a future ATLAS global infrastructure.

47
48 The remainder of the paper will proceed as follows. Section 2 will discuss the ATLAS
49 collaboration, as well as data movement requirements and expectations. Section 3 will discuss the
50 work to monitor and measure our networks. Section 4 will discuss the ATLAS effort to analyze our
51 network data. Section 5 will discuss PanDA and how it is evolving to better utilize the network.
52 Section 6 will cover exploratory work to determine the impact of future networks on ATLAS.
53 Section 7 concludes the paper.

54 **2. The ATLAS Collaboration**

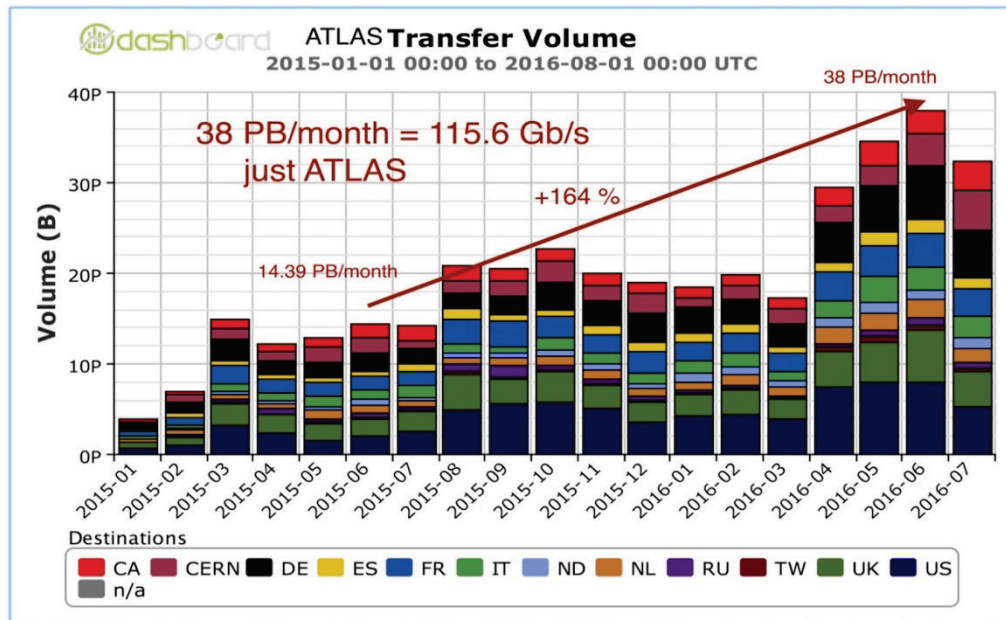
55 The ATLAS collaboration consists of over 3000 physicists and 1000 students from 38 countries and
56 178 Universities and Laboratories worldwide. This large group of scientists is working together at the
57 Large Hadron Collider (LHC) [1] to learn about the basic forces that have shaped our Universe since
58 the beginning of time and that will determine its fate. ATLAS physicists are exploring the frontiers of
59 high-energy physics in a number of ways: explaining the origin of mass, exploring the range of
60 validity of our standard model, searching for microscopic black holes, probing the existence of extra
61 dimensions, and looking for evidence of an as-of-yet undiscovered particle that may explain the dark
62 matter in our Universe.

63
64 To undertake these explorations, the collaboration has constructed the ATLAS detector [2] over a
65 15 year period and assembled it at Point 1 in the LHC ring. The detector is 45 meters long, 25 meters
66 high and weighs about 7000 tons. The ATLAS detector employs a number of types of sub-detectors to
67 measure attributes of the various particles resulting from the collision of counter-rotating beams of
68 protons in the LHC. There are millions of electronic channels associated with the readout of the
69 ATLAS detector. In effect, the set of all of these sub-detectors and associated readouts can be viewed
70 as a very large 3-dimensional digital camera, capable of taking precise "pictures" 40 million times a
71 second (proton beam bunches cross one another every 25 nanoseconds). The detailed information
72 collected allows the ATLAS physicists to reconstruct the underlying event and search for new physics.

73
74 If all the data ATLAS produces could be stored, it would fill more than 232,000 CDs per second, a
75 rate (and corresponding data-volume) which is not feasible to support with current technology.
76 Instead a set of hardware, firmware and software systems makes fast decisions about what data is
77 interesting to keep and result in a data rate of 400-1000 MBytes/sec into "offline" disk storage. Even
78 so, this rate of data production results in many petabytes of data being produced by ATLAS each year.
79 In addition, detailed simulations also produce Petabytes of data required to understand how the
80 ATLAS detector responds to various types of events and validate that the ATLAS software works as
81 expected. It is important to note that these large data volumes are common to all the LHC experiments
82 and not just ATLAS.

83 Because of the data-intensive nature of the ATLAS scientific program, the ATLAS collaboration
84 implicitly relies upon having a ubiquitous, high-performing, global network to enable its distributed
85 grid-computing infrastructure. Providing effective access to petabytes of data for thousands of
86 physicists all over the world just wouldn't be possible without the corresponding set of research and
87 education (R&E) networks that provide 1 to 10 to 100 Gigabits per second of bandwidth to enable
88 ATLAS data to flow to where it is needed. As can be seen in Figure 1, recent ATLAS wide-area

89 network use is rapidly increasing, continuing an exponential increase that has been observed since
 90 startup. This increasing use exemplifies the importance of networking for ATLAS and its globally
 91 distributed computing model.
 92



93
 94 **Figure 1 Recent ATLAS wide-area network use with a trend-line showing a 164% increase. ATLAS WAN use**
 95 **has grown to almost 38 Petabytes per month.**

96 Typical network paths that ATLAS data traverses consist of multiple administrative domains (local
 97 area networks at each end and possibly multiple campus, regional, national and international networks
 98 along the path). The ability of the Internet to allow these separate domains to transparently inter-
 99 operate is one of its greatest strengths. However, when a problem involving the network arises, that
 100 same transparency can make it very difficult to find the cause and location of the problem.
 101

102 Because of both the criticality of the network for ATLAS normal operations and the difficulty in
 103 identifying and locating the source of network problems when they occur, the US ATLAS facility
 104 began deploying and configuring perfSONAR-PS (now referred to simply as perfSONAR) in 2008.
 105 Our goal was to provide our sites with a set of tools and measurements that would allow them to
 106 differentiate network issues from end-site issues and to help localize and identify network specific
 107 problems to expedite their resolution. This effort evolved into first an ATLAS and eventually a WLCG
 108 [3] and OSG [4] effort to monitor and measure our networks.

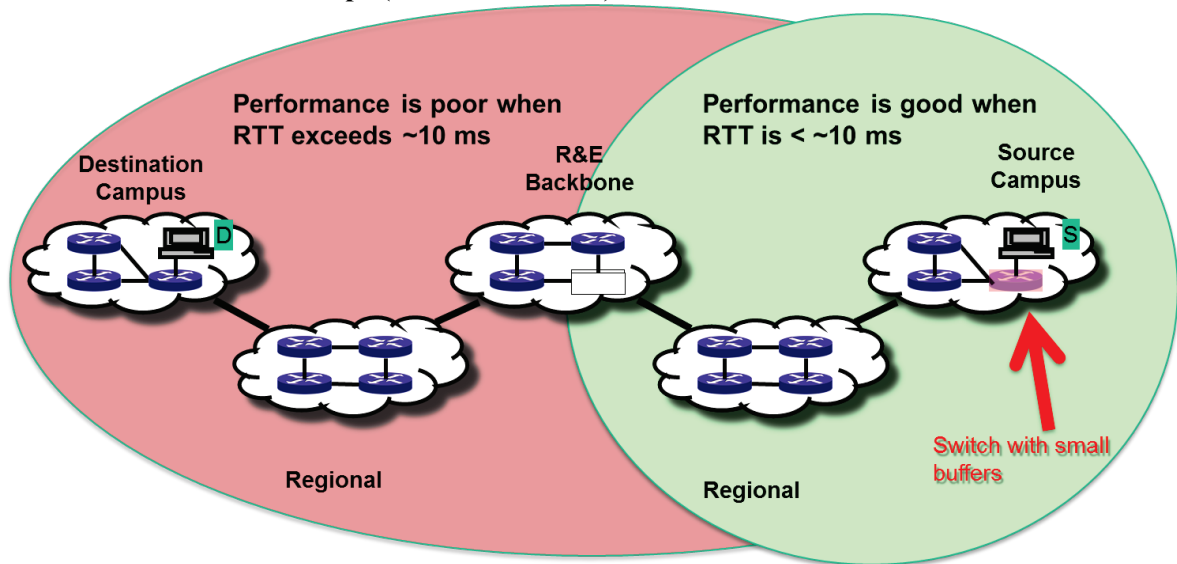
109 **3. Monitoring and Measurement our Networks**

110 Network problems can severely impact ATLAS’s workflows and have taken weeks or months to get
 111 addressed. End-to-end network issues are difficult to spot and localize because they are multi-domain
 112 (multiple independent administrators) and involve many components (end-systems, software,
 113 firmware, routers, switches, cables, etc). Standardizing on specific tools and methods allows ATLAS
 114 (and HEP in general) to focus resources more effectively and better self-support its collaborators.
 115 Thus we have chosen to use *perfSONAR*. *perfSONAR* is a framework that enables network
 116 performance information to be gathered and exchanged in a multi-domain, federated environment and
 117 its use in HEP was described in detail in a previous CHEP paper [5].
 118

119 Typical network problem involves packet-loss or packet reordering along a wide area network
 120 path. To illustrate the impact of packet loss on long network paths we can use the example shown in

121 Figure 2. Assuming the links shown are 10 Gbps, even a small loss can significantly impair the
 122 throughput. A 0.0046% loss (1 out of 22k packets) on 10G link results in very different throughput,
 123 depending upon the round-trip time (RTT):

- 124 • with **1ms RTT: 7.3 Gbps**
- 125 • with **51ms RTT: 122Mbps**
- 126 • with **88ms RTT: 60 Mbps** (factor 80 decrease)



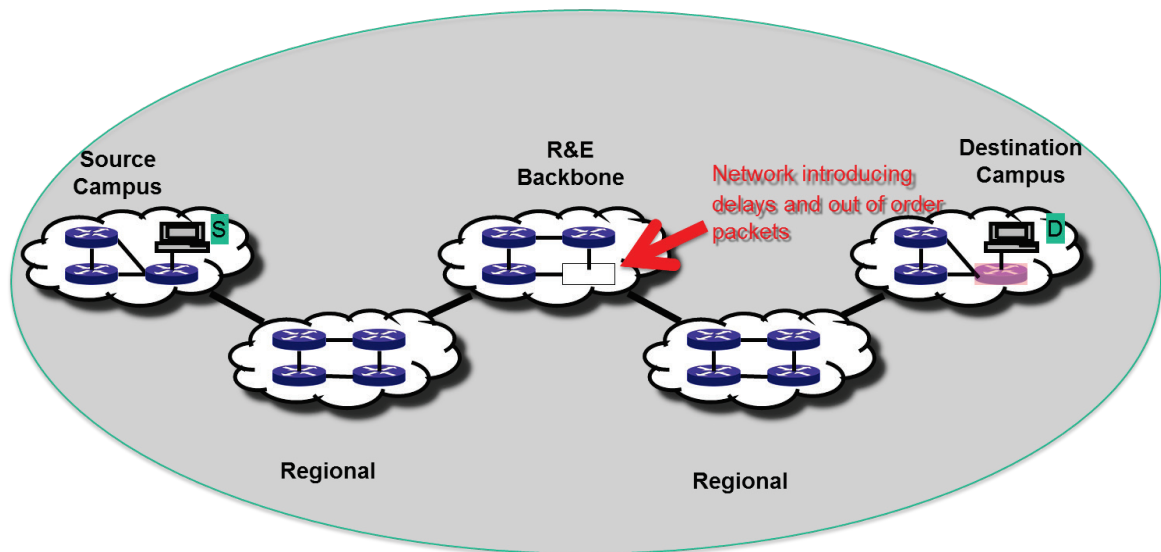
127
 128 **Figure 2 A wide area network path from a source of network traffic to a destination through many routers and**
 129 **switches. The behaviour of TCP in the presence of packet loss degrades significantly with round-trip time (RTT) and**
 130 **packet loss “close” to the source can mask network problems when traffic is to nearby destinations.**

131 The impact of packet reordering and jitter can also be significant. Referring to Figure 3 we note
 132 the impact of packet reordering when there is significant jitter. At 70ms RTT on 10 Gbps link, a 60
 133 second test results in significantly different throughput with only 1% packet reordering depending
 134 upon jitter:

- 135 • with 1% re-ordering, **0.2 ms jitter: 8.45 Gbps**
- 136 • with 1% re-ordering, **1.0 ms jitter: 1.10 Gbps**

137
 138 As we have seen, it is critical to understand when problems arise in the network adversely
 139 impacting ATLAS’s ability to use the network effectively. We rely upon *perfSONAR* to monitor and
 140 measure our networks. *perfSONAR* provides a number of standard metrics we use:

- 141 • **Latency** measurements provide one-way delays and packet loss metrics
 - 142 ○ Packet loss is almost always very bad for performance
- 143 • **Bandwidth** tests measure achievable throughput and track TCP retries (using Iperf3)
 - 144 ○ Provides a baseline to watch for changes; identify bottlenecks
- 145 • **Traceroute/Tracepath** track network topology
 - 146 ○ Measurements are only useful when we know the exact path they are taking through
 - 147 the network.
 - 148 ○ Tracepath additionally measures the Maximum Transmission Unit (MTU) on the end-
 - 149 to-end path but is frequently blocked from operating correctly because of incorrectly
 - 150 or over-zealously configured firewalls along the path.



151
 152 **Figure 3 A wide area network path from a source of traffic to a destination through links that introduce delays**
 153 **and packet reordering. In this case TCP throughput, even without packet loss, can be significantly degraded because**
 154 **of packet reordering and jitter (the variation in inter-packet arrival timing).**

155 **3.1. Organizing and Maintaining ATLAS Networking**

156 The ATLAS collaboration is benefiting-from and participating-in a number of efforts to instrument,
 157 measure, monitor, understand and control our networks. Since 2012 the Open Science Grid (OSG) has
 158 had a networking area whose goal is to provide network information and support to its members and
 159 collaborators. Since 2014 the WLCG has operated the Network and Transfer Metrics Working Group
 160 (NTMWG) which is responsible for instrumenting, measuring and reliably gathering *perfSONAR*
 161 network metrics from our networks. All ATLAS Tier-1 and Tier-2 centers are mandated to deploy
 162 *perfSONAR* Toolkit instances co-located with their storage resources. The original deployment
 163 campaign was led by the WLCG *perfSONAR* Deployment Task-force [6] which completed its work in
 164 2014.

165 Recently ATLAS has taken the lead in trying to analyze and better understand the network metrics
 166 being gathered by OSG and this work will be described in section 5. The ATLAS work is well aligned
 167 with the OSG goal of providing effective alarming and alerting for network problems.

168 One of the important activities of the NTMWG was to create a support unit [7] to coordinate
 169 responses to potential network issues. Tickets opened in the support group can be triaged to the right
 170 destination but networking experts from ATLAS or the other experiments. Many issues are potentially
 171 resolvable within the working group because of the information available from *perfSONAR*. More
 172 complex network issues can at least be identified and directed to the appropriate network support
 173 centers along with any additional supporting information. This has resulted in significantly
 174 decreasing the time it has taken to resolve network issues.

175 Lastly we should mention the LHC Optical Private Network (LHCOPN) and LHC Open Network
 176 Environment (LHCONE) [8] efforts. The LHCOPN was created in 2006 to implement, manage and
 177 maintain dedicated network circuits between CERN (the Tier-0) and the set of Tier-1 centers
 178 worldwide. A group of physicists, network engineers and members of global Research and Education
 179 (R&E) networks have met 2-3 times per year to manage and develop the LHCOPN since its inception.
 180 Because of the success of the LHCOPN in meeting the needs of ATLAS and the other LHC
 181 experiments, discussions were started concerning how this effort might be expanded to incorporate the
 182 needs of the Tier-2s and even Tier-3 computing sites worldwide. This discussion led to the formation
 183 of the LHCONE effort in 2012 and subsequently joint meetings with LHCOPN. The LHCOPN and
 184 LHCONE efforts are providing high-energy particle physics experiments like ATLAS with
 185 customized planning, services and development to support their global network requirements.

186 **4. ATLAS Analytics and Network Data Analysis**

187 The volume and complexity of the network metrics that are being gathered globally by OSG and
188 WLCG have created a pressing need to get this data into a location suitable for filtering and analysis.
189 The metrics are gathered along many paths across our global R&E networks and measure various
190 characteristics of those paths which change with time. To be able to more fully understand how our
191 networks are operating and especially to be able to identify and localize network problems, we need to
192 apply more complex analysis to our metrics than we can do with our existing *perfSONAR* toolkits
193 capability.

194 In late 2015 the OSG networking group began working with Ilija Vukotic, University of Chicago,
195 to feed network metrics into a new ELK (Elasticsearch, Logstash and Kibana) instance that was
196 already capturing many useful ATLAS metrics. Data being gathered by OSG from the global set of
197 WLCG-related *perfSONAR* instances was published to an ActiveMQ message bus instance hosted at
198 CERN and then sent to the Chicago ELK analytics instance via a customized FLUME instance. This
199 analytics service indexes historical network related data while providing predictive capabilities for
200 network throughput. Further details about this system are provided in this conference proceedings [9]

201 Using this analytics platform for network metrics was immediately valuable. By having all this
202 data query-able we were able to ask questions like: “Which sites have more than 2% packet loss to
203 more than 80% of their testing partners for the last 12 hours?” Being able to quickly find and localize
204 network problems is critical for our infrastructure performance. We are working on defining standard
205 alarm tables that continually update as new data is gathered that will serve as the basis of an eventual
206 alerting system.

207 One of the conclusions we were able to reach by having this network analytics capability is that
208 much of our ATLAS infrastructure is NOT tuned to take the best advantage of the networks we
209 currently have. There are a wide range of mis-configurations, non-optimal tunings and incorrect
210 application, firmware and hardware settings that lead to inefficient use of our networks. This wealth
211 of data now available and analyzable can identify bottlenecks and poor performance. We are now
212 working on the next steps to automatically and consistently find and fix such problems in ATLAS
213 resources.

214 **5. PanDA and ATLAS Workflow and the Network**

215 ATLAS relies upon the Production and Distributed Analysis (PanDA) workload management system
216 [10] to coordinate and optimize the collaboration’s set of tasks across it’s global resources. PanDA is
217 responsible for selecting the job execution site and it does this via a multi-level decision tree involving
218 task brokerage, job brokerage and a dispatcher. It also includes predictive workflows like the PD2P
219 (PanDA Dynamic Data Placement). Site selection was originally based upon processing and storage
220 requirements.

221 Recently PanDA has evolved [11] to incorporate network information as another component for
222 site selection because of the impact the network can have, both positively and negatively, on task
223 completion times and failure rates. The ATLAS analytics platform mentioned in section 4 is used to
224 summarize recent network metrics from FAX and *perfSONAR* and make it available for PanDA use.
225 This data augments other information PanDA already uses such as job completion metrics, errors and
226 timeouts per site.

227 The longer-term goal for PanDA is to go beyond network monitoring and treat the network as a
228 managed resource. Can we incorporate network provisioning, orchestration and control via software
229 defined network capabilities as part of PanDA? Initial simple tests have shown that network
230 knowledge is useful and beneficial for all phases of the job cycle. In both the ANSE and BigPanDA
231 projects we have added “hooks” into PanDA that could allow control of the network once production
232 quality mechanisms are in place to support that across at least some of our networks. This would be a
233 first and never attempted before for large scale automated WMS systems. To make this a reality for
234 ATLAS will require new, production quality capabilities from future networks.

235 **6. Exploring Future Networks for ATLAS**

236 Future networks won't just have increased capacity but are also enabling new interfaces and modes of
237 operation that can allow end-users to control some aspects of how data flows across networks. This is
238 referred to as software defined networking (SDN) and the primary impetus for these capabilities is
239 driven by commercial entities. In fact, SDN originated with Google and its attempt to better
240 orchestrate its data-center and wide-area networks.

241 An important question for ATLAS is whether or not SDN is something that can improve ATLAS's
242 ability to use its distributed resources. A group of people in the US from four of the ATLAS Tier-2
243 centers (AGLT2, MWT2, SWT2 and NET2) have begun to explore SDN ATLAS. The idea is that
244 we need a way to compare and contrast the impact of controlling the network versus using the network
245 as-is for real production ATLAS work. This means we need a non-disruptive way to incorporate
246 ATLAS production systems into an SDN testbed.

247 To do this we began working with the LHCONE point-to-point effort which has been exploring the
248 use of SDN to setup end-to-end network connections between LHCONE sites. The stumbling block
249 has always been getting SDN capabilities all the way to the source or sink of data. The R&E
250 networks may have different ways to setup circuits or control the "backbone" network but these never
251 reach into the local area networks (LANs) nor to the servers hosting the data. For the US ATLAS
252 sites, we proposed to solve this problem through the use of Open vSwitch [12].

253 With Open vSwitch (OVS) we have the ability to non-disruptively add SDN capability to existing
254 ATLAS production data servers at a few of our Tier-2 sites. This will give us the ability to selectively
255 test how ATLAS production behaves with and without SDN features in place.

256 Our plan is to deploy OVS on ATLAS production storage systems at all the participating Tier-2
257 sites as follows:

- 258 • Measure baseline performance on our systems for a few days.
- 259 • Install Open vSwitch v2.6.1 via RPM on all storage servers
- 260 • Reconfigure the network to move the server IP address from the Network Interface Card
261 (NIC) to the Open vSwitch virtual switch
- 262 • Verify continued normal operation of the system now that all network traffic is passing
263 through the vSwitch

264 Once we confirm that our systems continue to behave the same as before installing OVS, we can
265 proceed to test features of SDN between our participating sites and compare and contrast the site
266 performance when using these capabilities versus not using them. For example, one nice feature of
267 OVS is the ability to shape traffic by pacing the rate at which network packets are inserted in the
268 network. If we know storage systems are only capable of sourcing in sinking data at a certain
269 maximal rate, we can shape the corresponding network traffic to match that rate. This not only helps
270 reduce the load on the end-systems but also results in much better average performance across the
271 network based upon tests that have been conducted using OVS [13]. In addition there is very little
272 cost in terms of server resource use (roughly 1% additional CPU) to accurately shape traffic up to 100
273 Gbps. Finally, having OVS in place additionally allows various kinds of software defined network
274 controllers (e.g., OpenDaylight, Ryu, Floodlight) to see and interact with our servers, giving us the
275 possibility to orchestrate the network end-to-end for the first time.

276 We will need to do extensive testing once this capability is in place to understand the possible
277 benefits for ATLAS. One of the challenges involved is getting complete instructions for our various
278 end-sites on how to non-disruptively deploy OVS while those servers are in production. Assuming we
279 can successfully get this functioning with the US, we have requests from the Tier-1 sites in the
280 Netherlands and in Germany to also join in our testing. This will be important to test the possible
281 impact using the long fat network pipes across the Atlantic.

282 **7. Conclusion**

283 Networking is a critical component for ATLAS and underlies our distributed computing model.
284 Problems in the network can cause significant degradation for ATLAS workflows and can be very

285 hard to identify, locate and fix. To address this ATLAS has a working infrastructure in place to
286 monitor and measure our networks using *perfSONAR* and are benefitting-from and contributing-to
287 efforts around networking in OSG, WLCG and the LHCOPN and LHCONE communities. ATLAS is
288 also leading the effort to make complex analysis of network data possible and working towards new
289 capabilities in network notification and alerting, predictive network behavior and the identification of
290 problematic sites and servers. The ATLAS PanDA system is also evolving to take better advantage of
291 network knowledge and to prepare for future network capabilities that may allow control and
292 orchestration of our networks. Lastly, a group in ATLAS is exploring the possible impact of SDN and
293 testing how it may be able to benefit ATLAS.

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295
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298 contributors shown at <http://www.perfsonar.net/about/who-is-involved/>.

299
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