DAQExpert - An expert system to increase CMS data-taking efficiency

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Abstract. The efficiency of the Data Acquisition (DAQ) in the new DAQ system of the Compact Muon Solenoid (CMS) experiment for LHC Run-2 is constantly being improved. A significant factor on the data taking efficiency is the experience of the DAQ operator. One of the main responsibilities of DAQ operator is to carry out the proper recovery procedure in case of failure in data-taking. At the start of Run-2, understanding the problem and finding the right remedy could take a considerable amount of time, sometimes up to minutes. This was caused by the need to manually diagnose the error condition and to find the right recovery procedure out of an extended list which changed frequently over time. Operators heavily relied on the support of on-call experts, also outside working hours. Wrong decisions due to time pressure sometimes lead to an additional overhead in recovery time.

To increase the efficiency of CMS data-taking we developed a new expert system, the DAQExpert which provides shifters with optimal recovery suggestions instantly when the failure occurs. This tool significantly improves the response time of operators and the success rate of recovery procedures. Our goal is to cover all known failure conditions and to eventually trigger the recovery without human intervention wherever possible. This paper covers how we achieved two goals - making CMS more efficient and building a generic solution that can be used in other projects as well. More specifically we discuss how we: determine the optimal recovery suggestion, inject expert knowledge with minimum overhead, facilitate post-mortem analysis and reduce the amount of calls to on-call experts without deterioration of CMS efficiency. DAQExpert is a web application analyzing frequently updating monitoring data from all DAQ components and identifying problems based on expert knowledge expressed in small, independent logic-modules written in Java. Its results are presented in real-time in the control room via a web-based GUI and a sound-system in a form of short description of the current failure, and steps to recover. Additional features include SMS and e-mail notifications and statistical analysis based on reasoning output persisted in a relational database.

1. Introduction

The data-taking efficiency of Compact Muon Solenoid (CMS) is 95%. One of the major roles in the operations is played by the Data Acquisition (DAQ) system which controls the data-taking and selection of the events from the detector. There are a crew of operators who supervise the operations and a group of human experts who designed the detector and the DAQ system. Together they ensure smooth operations of the experiment and its high efficiency as a result.

Being able to act within seconds during operations of LHC is one of the challenges of the operator. The DAQ system is a complex, therefore there are tools to automate certain operations, e.g. starting the run, stopping, resetting, resyncing, configuring etc. Those tools facilitate the actions initiated by operators, however, it's not uncommon that the system goes

into erroneous state where deeper understanding of the system is required to overcome the problem in the optimal way. Experts of all subsystems prepare instructions for operators for specific problems. These are recovery instructions needed to resume the data-taking as soon as possible. Operators need to effectively select and issue them to make sure the operations of CMS go smoothly.

Over a time the list of recovery instructions grew. Even though it's prepared in an easy form of checklists helping to identify the problem, it takes considerable amount of time for operator to find the correct remedy when needed. The reaction time of the operator increased as it was difficult to remember all of them. Wrong or suboptimal decisions were not uncommon as well. This led to increased intervention time which can be divided into two parts: the reaction time (time between a problem occurrence and a subsequent recovery action taken) and the recovery time (time between a recovery action beginning and end).

Reducing the reaction time and selecting the optimal recovery action are the two areas for improvement that will reduce the overall intervention time and directly affect the efficiency of CMS. There are also other chores that this project aims to streamline. First is to facilitate the post mortem analysis for DAQ experts, another is to aggregate and to report a statistical analysis of the performance of the system and individual subsystems. Last but not least is to reduce the need of a external help especially outside of working hours without a deterioration of CMS efficiency.

1.1. Selecting optimal recovery action

The time spent recovering is the main contributor to the overall intervention time in CMS and 61 amounts to 82.5%. Therefore its essential to define and select the quickest set of actions needed 62 to bring the system back to operation. Many different recovery procedures may be distinguished, 63 e.g.: stop the run and start, reconfigure specific subsystem, red and green recycle, resync, restart. 64 They differ significantly in terms of impact they have on system and how much time they take 65 to complete. Some of them take few seconds whereas other few minutes. One of observed habits 66 is issuing by the operator the most versatile and heavy recovery action for most problems. It 67 gives the highest chance to recover but in most cases it means an overhead in recovery time. 68 The aim of the project is to help operator make optimal decisions. 69

o 1.2. Reaction time

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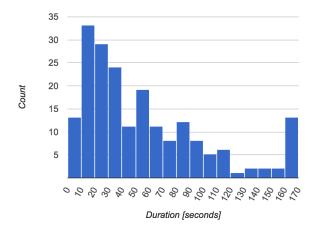
Time from the problem occurrence to the first recovery action is a second area to improve.
There are sound notifications in the control room to attract attention of the shifter in case data
flow stops. Identification of the problem and selecting the recovery action takes from couple of
seconds up to minutes. Figure 1 presents the histogram of a reaction time over the period of
6 months of data-taking (Aug 2016 - Jul 2017, excluding YETS - Dec 2016 - May 2016). Most
of the times operator reacted in the time 10-20 seconds - 33 occurrences. The aim of the project
is to reduce the reaction time and as a result move the distribution to the left hand side.

1.3. Minimizing external help

In case the operators cannot fix the problem themselves they can always rely on help of DAQ on-call experts. They are available 24/7. As much as they are helpful, they are not fans of night calls especially with trivial problems. Minimizing the number of these calls especially outside of working hours without a deterioration of CMS efficiency is one of the goals of the project.

1.4. Facilitating post mortem analysis

After the problem occurs and is addressed, usually there is a lesson to be learnt from it to avoid similar problems in the future. Numerous monitoring systems enable the access to data



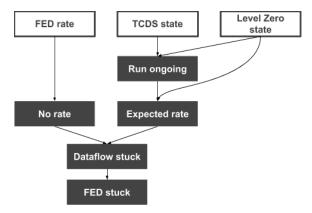


Figure 1. Reaction time histogram.

Figure 2. Subset of Logic Module hierarchy.

necessary to analyze the problem but not without the flaws. Some data is available after the problem has been resolved, some only during the problem is ongoing. The aim of the project is also to streamline the post mortem analysis by making both historic and real-time data easily available in the same way at any moment.

2. Expert system and architecture

There were particular external forces that influenced the design of the project and technologies used. Specifically no rule-based engine framework has been used to implement reasoning and define expert knowledge.

Three services has been distinguished. Firstly there is one that aggregates all monitoring information into a snapshot and persists it - Snapshot Service. Secondly there is one that applies the knowledge of the experts on the snapshots - Reasoning Service. Finally there is a service that manages and delivers notifications, including sms, e-mail and live suggestions and sound notifications in control room - Notification Manager. The DAQExpert brings all parts together in a form of web application allowing to browse whole history of monitoring data, to view the analysis, to control the snapshotting of a system and to manage notifications.

2.1. External forces

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It is highly likely that project's lifetime, like many other ones at CERN, will be more than a decade and will be maintained by different people over time. Thus, one of the main constraints was to avoid short-lived and highly specific technologies that would make it difficult to new team members to quickly become productive. It needs to be easy to understand and modify requiring little effort to learn. Rule-based engines have beed evaluated in the past by the CMS DAQ group and were considered not suitable due to steep learning curve. Introducing additional language was not an option. The solution needs to be based on current skill set of the team members and as simple as possible.

Moreover, the nature of the project foresees frequent logic-related updates possibly introduced by many authors. If there is new recovery solution advised by sub-system experts the according knowledge needs to be introduced to the system. Finally a high availability needs to be assured, as each downtime of the system means no support for the operators and possibly extended downtime of CMS.

As a result the new expert framework has been designed from ground up. It allows to define the knowledge in imperative language popular in the group - Java. The knowledge is organized in small independent Logic Modules which are building blocks of the system. 2.2. Monitoring, aggregation and data flow

There are multiple steps before information from data sources is available to the clients. First the monitoring data sources are queried to get information about data taking health. There are multiple heterogenous sources of monitoring data. They differ not only in terms of what data is available but also when and how quickly the data can be retrieved. Diagnosing the problems in data-flow in real time requires quick access to key information. Post mortem analysis requires access to all relevant data at any time. To enable both the so-called snapshot of the system is taken periodically. It contains all necessary information to identify and understand the problem. Thus, the first step was to aggregate all the information in snapshots that brings both structural data and instantaneous state of all components. The scope of the snapshot has been designed to give full picture of system state at given moment. Both real-time and post-mortem analysis are based on the same data from snapshots.

While the snapshot is persisted for possible later post-mortem analysis it is simultaneously sent to *Reasoning Service* for real time analysis. As a result of analysis, notifications may be yielded that will be later dispatched by the *Notification Service* and delivered to the clients. On the way, all of the services persist their results enabling to browse historic data.

134 2.3. Knowledge definition

Its common in the industry that the expert knowledge is defined in a declarative language. For instance in a form of if-then-else rules expressed in a high-level rules language. As a result of external forces this way of defining knowledge has not been used. The new custom made framework based on *Logic Modules* has been introduced. A *Logic Module* (hereinafter LM) is a building block of *DAQExpert* expert logic. Each LM focuses on one thing: expressing piece of knowledge about DAQ system:

- each LM defines one condition,
- the definition of condition is placed in satisfy method,
- the method returns true if condition is satisfied and false otherwise,
 - one LM can use results of another LM.

The knowledge defined in LMs is used to find optimal solution when problem occurs in dataflow. The set of LM is defined by DAQ experts. LMs operate on the snapshot and verify if key parameters are in expected states or in specified value ranges. As LMs may use outputs of other LMs there is a predefined order of firing them. Figure 2 presents subset of LMs responsible for identifying one of many problems that may occur during data-taking. In the example there are five LMs and three parameters being monitored. The LMs are fired in top-down order allowing LMs rely on others as arrows indicate.

LM may play two roles: it might be a sub-step of the analysis (*Expected rate*, *No rate*, *Dataflow stuck*) and final step of the analysis (*FED stuck*) delivering description of the condition and recovery suggestion. Finding optimal recovery to given condition in this hierarchy means that the chain of LM was activated:

- Run ongoing there was a run started according to TCDS state and LevelZero state,
- No rate the summarized FED rate was equal to 0, there was no data flowing
- Expected rate the TCDS state and LevelZero state indicate that no recovery action is being issued right now, all subsystems are running and the rate is expected to be non-zero, note that this LM is also checking if there is currently a run ongoing using output of Run ongoing LM
- Dataflow stuck this summarizes outputs of two LMs: No rate and Expected rate, it's activated when two related LMs are active meaning that the data-flow is stuck,

• FED stuck - starts the analysis when there is Dataflow stuck, it performs specific checks that reveal the specific FED to be stuck, causing the data-flow to be stuck, this LM identifies precisely the problem thus it consists final instructions for the operator including detailed description of the problem and recovery suggestion.

3. Results

The current version of the service (2.9.0) already improves all of the areas described in the first section. Operators of the DAQ system at CMS have now at their disposal a tool delivering recovery action suggestions in the real-time - the Dashboard (see figure 3). Whenever the problem occurs the tool is showing the description of the problem and the optimal steps to recover. It helps shifters to take right decisions, reducing both reaction time and need of external human expert help.

Measuring the success of the project was quite challenging task. The ultimate goal is to increase the efficiency of data-taking at the CMS. However the overall efficiency is not an adequate parameter as there are many different factors that interfere with DAQExpert. The detector is constantly being improved, there were major upgrades in both hardware and software that influenced the final efficiency of the CMS.

The feedback of operators and human experts is important as they are the target user groups of the system. As much as they appreciate the new system, the verbal feedback cannot be used as a measure of success as it is not quantitive and it is subjective.

Three parameters have been identified to adequately represent the success of the project: accuracy of the recovery action selection, reaction time of the operators and demand of the external help. They are practically independent of other factors which make them reliable means of comparison.

Expert was introduced in mid 2016 but operators were instructed to use it in the early 2017. All comparisons are based on 6 months of data-taking (August 2016 - July 2017, excluding Year End Technical Stop (YETS) that lasted December 2016 - April 2017, and May 2017). Months December of 2016 and May of 2017 were excluded due to special conditions of data-taking and the fact that the operations lasted only for few days. This gives four months from 2016 where there was no *DAQExpert* support and two months from 2017 where the support was available.

3.1. Recovery selection correctness

The improvement was mostly observed in the number of optimal recovery decisions taken by operators in case of data-flow upsets. As stated before the recovery time is major part of total intervention time (82.5%), thus it is essential to minimize it by choosing the most appropriate and time-efficient recovery actions (optimal recovery actions).

There are two main categories of recovery actions: heavy - lasting usually for 3-4 minutes, initiated by e.g. stopping the run; and light - lasting 7-10 seconds initiated by e.g. resetting the system. Suboptimal decisions mean unnecessary time spent on recovery or irrelevant action being performed that will not solve the problem leading to same problematic condition after its completion. On average it means that each suboptimal decision adds roughly 93-125 seconds of overhead to recovery time.

In 2016, during data-taking, operators had no guidance and there was 36 data-taking failures, 32 times the optimal decision was taken, 4 times the decision was not good. In 2017 there were 29 data-taking failures and all of the decisions were optimal.

3.2. Reaction time

This may not be the most important area to improve as it is minor part of total intervention time - 17.5%. There is high variation depending on individual operator capabilities. The reaction time is now slightly lower. The average reaction time in 2016 is 67 seconds while in 2017 it is

60. It's an area where human factor plays great role and bypassing the operators will certainly assure better results.

3.3. Non-quantitative results

An improvement in the field of post mortem analysis has been made as well. In case of more demanding data-flow upsets where further investigation is required the *DAQExpert* streamlines the chores of human experts. First of all the go-back in time functions has been introduced that enable human experts to access the historic data easily with the tool *Browser*, see on figure 4. Second of all there is more data available to support post-mortem analysis than ever before. All parameters of the system are kept for some period of time when debugging demand is more likely. Later on the resolution of data is reduced so that the disk space requirements are feasible.



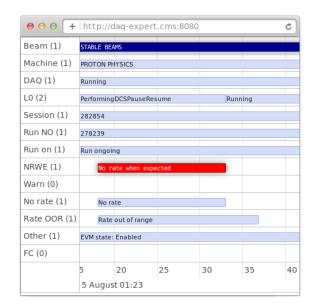


Figure 3. Dashboard - tool to show suggestions in real time.

Figure 4. Browser - tool to browse in time to enable post mortem analysis.

3.4. Summary

Both quantitative and non-quantitative results were presented in this chapter. The recovery action selection accuracy was improved by introducing the expert system. There were no suboptimal recovery decisions observed anymore, whereas before they happened 11.2% of time. Additionally the reaction time shrunk by 10.4%, from 67 to 60 seconds. It means that total intervention time, including reaction time and recovery procedure time is now shorter by at least 10.3 seconds on average. One major improvement is to bypass the operator whenever possible. This will shrink the reaction time to milliseconds. External help demand was not quantified yet. This area is of great interest for on-call human experts as it has a potential of limiting disadvantages of their obligatory shifts. Number of calls (normalized by number of problems to on-call before and after DAQExpert introduction may reveal trends and influence of the system.

References

[1] CMS Collaboration, JINST 3 S08004 (2008)