

# A FLEXIBLE MOBILE NETWORK MONITORING TOOL

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Abstract: This article presents a model to implement an information system to monitor mobile networks using Key Performance Indicators (KPIs). The proposed model encloses not only the KPI calculations but the translation, importing and presentation of data. It also addresses more advanced topics such as system configuration, node dependencies and routing. A description of the system's implementation is also presented, from two perspectives: high-level components and technology-related issues. The paper finishes with the presentation of a system application: monitoring a KPI using a data plot generated by the system.

## 1 INTRODUCTION

Designing an information system to monitor the performance of a mobile network presents quite interesting challenges. The inherently complex network structure faces the implementer with issues regarding data conversion, storage, processing and presentation.

Network and node performance measurement is crucial for any telecommunications network. It allows network scanning and spotting of anomalous situations that may contribute to a decrease in the required Quality of Service (QoS) (Kreher, 2006). With the help of an appropriate tool to diagnose these errors, one can apply corrective measures that eliminate or mitigate the problem's causes and effects.

Performance measures are usually achieved by collecting statistical data at the nodes and then processing it with appropriate mathematical formulas to obtain the required performance issues. This leads to the concept of Key Performance Indicator (KPI). A KPI is a formula-based computation executed over collected data. Usually, the formulas that describe KPIs are simple algebraic equations. The difficult issues that may involve KPI's calculation are related to the information filtering and data selection that takes place beforehand (Kreher, 2006).

In mobile networks the data is usually collected and transmitted to a central repository where it is stored. For example, in a typical Global System for Mobile Communications (GSM) network this would be the Operation and Support Subsystem (OSS).

It is common to find different formats to encode and transmit the nodes collected data. This discrepancy is due to the continuous network evolution and expansion which leads to the introduction of different technologies at the nodes.

The next step is to process the data and that is where KPIs are evaluated. Their formulas are applied on the raw statistical data and the results are again stored on a different data repository close to the network operator equipment. Therefore operators, or subcontracted network managing service providers, must carefully dimension its database capacity and structure to implement a Network Monitoring Information System (NMIS). Finally the processed data must be presented to the user in a convenient way.

In this paper we describe a NMIS model that was completely developed and tested to fulfil an operator's needs.

Our goal is to develop a NMIS model that can be easily customized and with great adaptability capabilities. Our system must hide all the network

monitoring complexity, be robust and allow distributed access.

The remainder of this paper is organized as follows: section 2 briefly describes the state of the art on these system tools, section 3 presents the system model, as well as a description of the model's implementation, section 4 presents the results, based on a network monitoring service example and it briefly discusses the impact of this model on the achievement of the service goal. Finally, section 5 summarizes the model conclusions.

## 2 STATE OF THE ART

Since network monitoring is not a common issue on the software developing circles, there are not many solutions that can serve as a *de facto* standard to base a new system development. Although we were able to find some similar solutions, these are commercial products and therefore their implementation specifications are not in the public domain.

Nevertheless, Waadt et al. (Waadt et al, 2005) presented a framework for a QoS monitoring, alerting and SMS Center reconfiguration that we can use as a model for the features that we have developed. Their system is written in Java and operates in real-time. It detects when the QoS parameters drop below, or exceeds previously defined thresholds, and generates an alert message that can be used for system reconfiguration. This system is strongly biased towards the Short Message System (SMS) of GSM networks and thus it does not represent a full solution for a more general performance monitoring system.

Rigallo et al. (Rigallo et al, 2002) have also proposed an architecture and functional implementation of a real-time monitoring and operation assistant system developed for Telecom Italia Mobile (TIM). This system, named NetDoctor, is a multi-layer OSS that allows for data collection and storage, network status monitoring and automatic failure detection and correction. On top of this, there is a web interface that allows distributed access.

Both proposals come short when the goals of tool design encompass a broad scope solution covering different areas of mobile networks while maintaining a good degree of simplicity. These solutions are based on frameworks, which, as it can be found on (Fayad et al, 1999), impose a steep learning curve on their users. We have deviated from this option by developing a "service-oriented" solution where the services can be thought of as

components, thus achieving the desired simplicity, ease of use and also rapid-development ability as a side effect.

## 3 SYSTEM MODEL AND IMPLEMENTATION

In the following description it is useful to separate the system into two major components: core services and auxiliary services.

Core services deal with storage, processing and presentation capabilities. Auxiliary services are responsible for filling the gap between the physical network and the data model used in the core services. This insulating layer must accept the different data formats used by the network nodes and convert them to a standard system format. It must also import data into the storage technology used by the core services. The auxiliary services act as an interface between the real world and the core services.

On the other hand, core services handle the system's main concerns: they store the raw and processed data in a structured way to allow great system's flexibility with the chosen database technology, they supply a mechanism to compute the KPIs from the defined formulas and they enable data to be presented in an informative, distributed and user-friendly way.

The following figure illustrates the system modelling in terms of components and its relations with the real world.

The storage service, which is part of the system's core services, encompasses the system data model and the repository where those data is stored. This repository is traditionally implemented as a relational database system (Codd, 1970). The system's data model can be easily driven by analysing the proper network hierarchy:

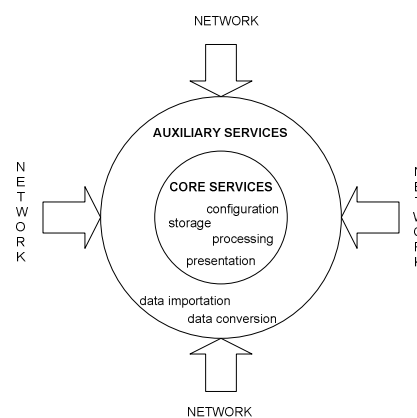


Figure 1: Modelling the system in layers.

“A network is a set of nodes. A node belongs to a certain type of network element. This has raw statistical data objects and KPIs associated to it. A node can depend on another node, and can have zero or many nodes that depend on it. A node can be connected to one or more nodes and this connection is named a route.”

Using these relations we can represent the data model structure using a Unified Modelling Language (UML) (Booch et al, 1970) class diagram. Figure 2 shows a simplified class diagram that might be helpful to provide better understanding of the system’s data model.

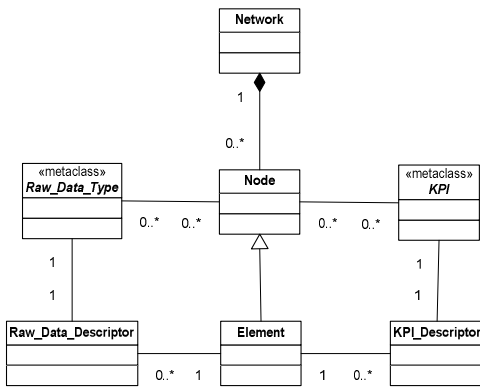


Figure 2: Simplified class diagram of the main system.

We present the statistical data (raw and KPIs) on different tables, with one table per KPI/statistical data object. We must keep record of the data objects attributes and values and this leads us to the concepts of “descriptor” – an entity that is responsible to describe other entities in terms of their attributes – and “value holder” – an entity that keeps the value of another entity defined by a descriptor. So we should have a table to keep information about data items such as name, formula and network element to which the data is associated, and another one to hold the corresponding data values. The first table is the “descriptor” and the second one is the “value holder”. This distinction makes sense from the database point of view as it increases its efficiency and provides a clean structure. The <<metaclass>> UML stereotype is used to indicate that both raw data types and KPIs will have a similar structure, but different table names and column number and names.

Figure 3 illustrates a class diagram showing node dependency and routing. A node can act as a parent for other nodes, or a as child of one node. Therefore, a parent-node can be parent of zero or more nodes, and a child-node can have, at most, one parent. A route can also be modelled as an association

between two nodes. The roles of a node in a route are origin and destination.

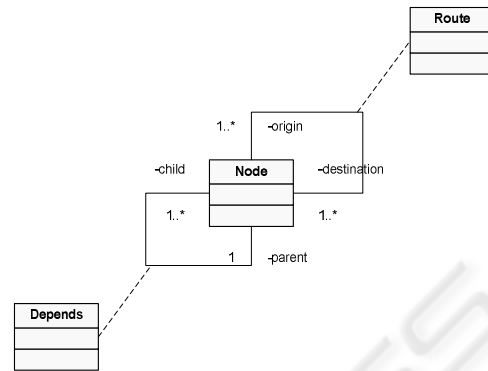


Figure 3: Simplified class diagram showing the routes and node dependency.

With this static structure we can implement the storage services of the system. Processing services, as previously mentioned, are responsible for taking the raw statistical data, applying the appropriate formulas and calculating the corresponding KPI values. Although this task can be accomplished in several ways, it will always have to follow these stages: retrieve the formula from the KPI descriptor, retrieve the data from the raw data value holder, perform the calculation and store the result on the KPI value-holder.

Presentation services are responsible for fetching data (both raw and KPI) and organizing it in a convenient format for a human reader. They can also be implemented in several different ways.

Configuration services are used to manage networks, element types and nodes, as well as to define KPIs and raw data statistical objects.

Another type of services, called auxiliary, must be able to translate the incoming data to a format that is understood by the system and then store it in the database. This translation process is divided in two stages: parsing the input data and then processing it to the output format. The next step is to import the output data into the database. However, this depends on the technology used, according to its database management system. To implement these tasks, one must know the formats for input data and structure of the database tables, where parsed data will be stored.

To validate our model, we made a complete system’s implementation, fed it with real network data and then we ran some tests on it. To undertake this task we have implemented the core part of a GSM mobile network.

The implementation was done on a Linux machine with MySQL database system, a PHP 5 interpreter and an Apache http server. This

configuration is widely known as LAMP-bundle and was chosen due to two criteria: proven system stability and zero cost.

We propose a model sufficiently general to accommodate any kind of network whose nodes may be associated with statistical measures. Another feature present in our model is portability, allowing it to be easily migrated to different platforms with low production costs. That is why we chose PHP, an interpreted language, and MySQL. PHP also has the advantages of strong scalability capacities and a large set of libraries that allow the programmer to perform many complex tasks without having to rewrite code for them. Furthermore it can run both as a server-side application and as a command-line interface (CLI) application which allows the implementation of all systems services in PHP. In fact, the data parsers, the data import-mechanism, the calculus engine used by the processing services and all the web content generation code were coded using PHP. The fact of having only one programming language in the system also benefits code maintenance.

Parsers convert statistical data received from nodes – via the OSS – in a convenient and meaningful format for the system. We chose this data format as Delimiter-Separated Values (DSV) because it is well suited to be imported in MySQL database systems. For convenience, these parsers output files whose name identify the node that produces the data, the raw data type name and the date of collection. The parsing process is based on regular expressions matching,

After the parsing process, the output data is stored in the database. For this we wrote a script that reads the processed data files and tries to match its meta-information (file name and headers, if applicable) with the raw data descriptors stored in the database. When a match is found, the data is retrieved from that file, line by line, and written on the corresponding database table.

These parsing and import processes were scheduled as cron jobs. Cron is a time-based scheduler application found on Unix-like systems. Since these might be resource-consuming activities they were scheduled to run on an off-peak hour (3 A.M.).

After storage is completed, another cron job executes the processing services. These were implemented as a software engine with the ability to perform mathematical calculus operations. For this purpose we chose a very useful PHP function: *eval()*. This function receives a string parameter and evaluates it as if it was code. It is useful for situations where one wants to store code on a database and execute it later on, which is exactly the case we are interested in. The processing service

retrieves the KPI formulas on the descriptor table of the database and evaluates them, for the different raw data samples, using the *eval()* function. After evaluation the results are again stored on the appropriate KPI tables on the database.

Presentation services are also considered core services because this system is user-oriented, so the interface between the data and the user plays a crucial role in the system. These services were implemented as a web-site where the user can request information on several KPIs. A plot for a given KPI, a set of nodes and date interval was one of the implemented services. This service allows a user to follow the variation of a given KPI. To generate the plots we have used a PHP library named 'jppgraph', which is also open-source. Other services were also implemented by operator request.

Configuration services were also implemented as services running on a web-site. Using Hypertext Mark-up Language (HTML) forms it is possible to obtain a clear and intuitive way of configuring all the system's parameters regarding its management.

## 4 RESULTS AND DISCUSSION

For illustrating the results obtained with our system we present in figures 4 and 5 the results produced by a KPI named "CP LOAD". This shows the CPU load of a MSC node and it is evaluated with the following formula:

$$CP\_LOAD = ACCLOAD / NSCANS \quad (1)$$

Where ACCLOAD is the accumulated processor load in percentage and NSCANS is the number of accumulations. These fields were implemented as counters on the raw data object that comes from that node.

In figure 4 we present the figure obtained for the evolution of KPI "CP LOAD" on a MSC located at Luanda, Angola, between 15th and 28th December 2006. It is possible to see a recurring pattern every couple of days, as well as to determine the lower and higher mean value intervals. The x-axis represents time and the y-axis the KPI's percentual value.

As we can observe in Figure 4, there was no data for 18<sup>th</sup> December and also between 23<sup>rd</sup> and 24<sup>th</sup> December.

Figure 5 shows a different view for the same KPI and a given date, Christmas Day on 2006, now concerning two different MSCs located at Luanda and Benguela, Angola. This figure clearly shows the traffic evolution on that day, making it easy to figure out its busy-hour.

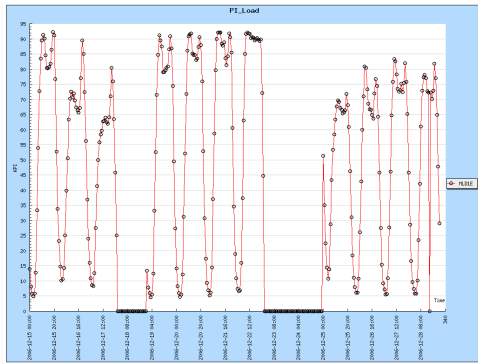


Figure 4: Evolution of “CP LOAD” between 15<sup>th</sup> December 2006 and 28<sup>th</sup> December 2006.

These results, as well as the validation of our complete system, were only possible thanks to an operator that has provided us the necessary data.

### 5 CONCLUSIONS

In this paper we have proposed a system model to implement a network monitoring tool that can be used to manage networks from telecommunication operators or other systems alike. This model is highly flexible and can be easily extended to accommodate any particular operator’s needs. The main characteristics of this system are its low development costs (due to the use of open source software), portability (since all code is written in an interpreted language, and thus independent from the system), reduced complexity (as opposed to the large framework-based software referred on the state-of-the-art), modularity (its services are independent modules), flexibility (a consequence of its simple structure and modularity) and easy remote access (via a web page).

The system has been implemented, validated and tested in a real environment of a GSM operator.

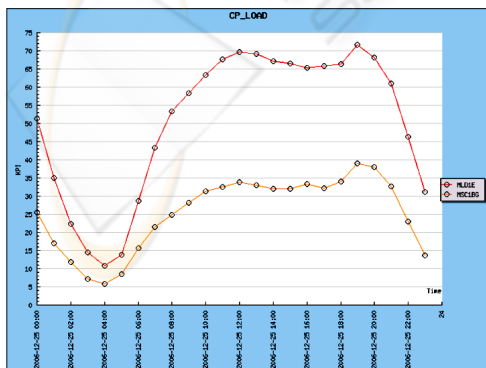


Figure 5: Evolution of “CP LOAD” on two different MSCs on 25<sup>th</sup> December 2006.

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