

Size Measures for Large Web Service Systems

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Abstract: Web service systems grow larger with age whenever organizations add new services to existing systems. As is the case with other types of software, very large Web service systems are difficult to understand and maintain and are therefore undesirable. A couple of measures have been proposed in literature that can be used to analyze the size attribute of Web service systems with the goal of aiding designers and managers in the management of such software. However, these measures target only simple to medium-sized services, and are not effective for very large cross-enterprise services. In this paper, we propose some size measures for evaluating the size of Web service systems irrespective of their granularity, thereby providing useful information to business process managers. We have validated the measures theoretically using Briand's measurement framework.

1 INTRODUCTION

Web service systems are an important tool that enables interoperability between today's Web-based organizations that need to conduct business transactions with their partners across different platforms. Several Web services composition languages have been proposed such as the popular Business Process Execution Language (BPEL) (Modafferi and Conforti, 2006; Zheng et al., 2007). BPEL is built on top of the Web Services Description Language (WSDL) and therefore, all BPEL processes are also implemented as services (Michelson, 2005). The research work presented in this paper is applicable to all orchestration-based Web service systems such as those implemented with BPEL language.

Due to Web services composition routines where new services are added each time new functionality is needed, the resulting systems can grow very large with age (Cardoso, 2008). Many researchers agree that very large systems are difficult to understand and to maintain (Cardoso, 2008; Munoz et al., 2010; Rolon et al., 2008). Furthermore, existing language technologies such as BPEL are ill equipped to manage very large Web service systems due to their lack of flexibility and lack of strong modularity features (Charfi and Mezini, 2004). Managers of

Web service systems created with such languages are therefore very much concerned about their quality.

In an effort to address the above issues, several authors have proposed a measurement-based solution. Software measures provide managers with information on potentially risky systems, which in turn helps them to make a decision on what to do with such systems. Some of the size measures that have been proposed either in the Web services or business process area include the number of activities (NOA) (Cardoso et al., 2006; Gruhn and Laue, 2006), number of basic activities (NOBA) and number of structured activities (NOSA) (Muketha et al., 2010), and number of nodes in a graph (Mendling and Neumann, 2007). While these measures are good for evaluating simple to medium-sized services that are atomic in nature, they are inadequate for measuring Web service systems that span across several enterprises and that consist of several interacting atomic services. There is a need for scalable size measures that can measure the size of Web service systems irrespective of their granularity.

In this paper, we propose a size measure for large Web service systems. The measure implements a simple approach where low level services (simple services) are measured first, and then a summation of their separate values is computed as the size of

medium level services. The high level measure is computed in a like manner i.e. as a summation of all the values of the services at lower levels. The proposed measure has been validated theoretically using Briand’s generic measurement framework (Briand et al., 1996) and the results are presented.

The rest of this paper is organized as follows. Section 2 presents related work, Section 3 describes Web service systems, Section 4 presents the proposed measures, Section 5 presents results, and Section 6 presents the conclusions.

2 RELATED WORK

A couple of size measures have been proposed that can be used for Web services systems.

Cardoso et al. (2006) and Gruhn and Laue (2006) have separately proposed the number of activities (NOA) measure as a business process equivalent of the lines of code size metric for software. Other similar measures proposed by Cardoso et al. (2006) include the number of activities and control-flows and the number of activities, joins and splits.

In (Muketha et al., 2010), two size measures for business processes called number of basic activities (NOBA) and number of structured activities (NOSA) are also proposed.

Another related size measure that is relevant to Web service systems such as number of services (NS) may also be found in (Rud et al., 2006) and (Zhang and Li, 2009).

The NS can be useful in choreography-based systems, but may be inadequate where orchestration-based systems are involved. Other measures mentioned in this section are limited in that they target only simple to medium-sized atomic services, but not large Web service systems.

3 WEB SERVICE SYSTEMS

Web service systems grow naturally out of a need by organizations to add more functionality to existing systems. Two main architectures for building Web service systems are orchestration and choreography. In both cases, several related services are composed into a larger system.

Choreography refers to the conversations between the various Web services in a peer-to-peer style while orchestration refers to one service being designated as the controlling service. Orchestration is an environment where a controlling service

invokes all other services needed in order to execute a business function for a specified customer. A detailed discussion on these two architectures can be found in the work of Daniel and Pernici (2006).

According to Cardoso (2008), Web services are simple applications performing one function. However, this definition is insufficient to shed light on exactly what a Web service is or even what its capabilities are. Several types of services have been identified based on their functionality. For instance, Michelson (1995) states that services may take the form of request/reply, worker, agent, aggregator, or a process. Table 1 describes these types of services that may be found in a typical orchestration-based Web services system.

Table 1: Types of services found in an orchestration-based Web services system.

Service	Description
Request/reply	Retrieves information (but may also modify the information) before forwarding the result to the requestor
Worker	Performs specific function (e.g. calculation)
Monitor	Observes something and then gives a report on its findings based on some monitoring rules
Agent	Similar to a monitor (i.e. observes something based on some monitoring rules). In addition, it acts on its findings.
Intermediary	Intercepts a service a message and then performs a value-added function on it after which it forwards it to its original intended destination
Aggregator	Combines results from other services
Process	A long-running service that controls other services needed to fulfil a particular business goal.

As mentioned earlier, orchestration-based service systems always have one controlling service (CS) designated as the process. Such systems are long-grained, and have been known to grow quite large and complex with time as new services are added to the existing system. As is the case with other types of software systems, high complexity affects the external quality of the system, something that is undesirable. In this paper, we identify three granularity levels of Web service systems. These include:

- Low granularity level
- Medium granularity level
- High granularity level

The low granularity level consists of atomic instances of services as described in Table 1. These services may be hosted in a single node or they may be residing in different nodes. In addition, these services might have been created using several different programming languages, especially the workers that may be required to perform certain functions such as math calculations. Since our focus is on BPEL services systems, the implementation details of atomic service within each system are therefore transparent.

At the medium granularity level, the Web service system may cut across an enterprise. All types of services described in Table 1 might be present. However, these services are seen as elements of one large Web service system, and interacting together. One of the services is designated as the controlling service (CS), also called process, which means it has the logic needed to control all interactions that may be required by the system.

Figure 1 shows a medium granularity service (MGS). Figure 1(a) represents the process, and Figure 1(b) represents a monitor service. In Figure 1, white-shaded circles represent activity within a service, directed arrows indicate internal control-flow within a service, and directed dash-line arrows indicate control-flow between one service and another.

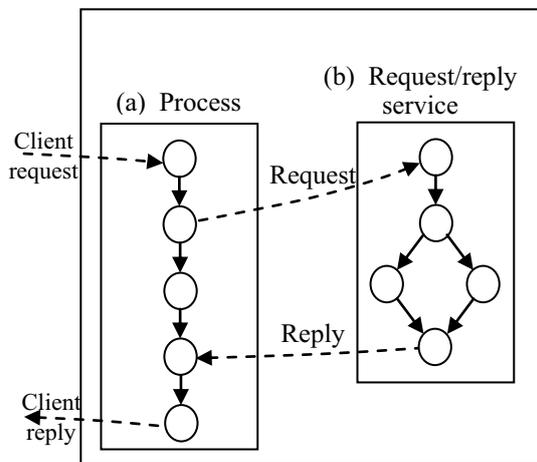


Figure 1: A system model representing a medium granularity service (MGS).

High granularity services (HGS) are very large systems that may span across several enterprises. These systems have a similar architecture to those at the medium level, except that they are colossal in nature. This means that quality challenges are much higher here than those in the smaller systems. To illustrate this point, the CS of a HGS treats MGS as

compound services and therefore invokes them alongside the regular atomic services. Furthermore, the depth of invocations might go to several levels. The example in Figure 2 shows a HGS system containing four interacting services: a process, two medium granularity services, and one low granularity service (a monitor service). For simplicity purposes, all interacting services are represented as rectangles in the figure.

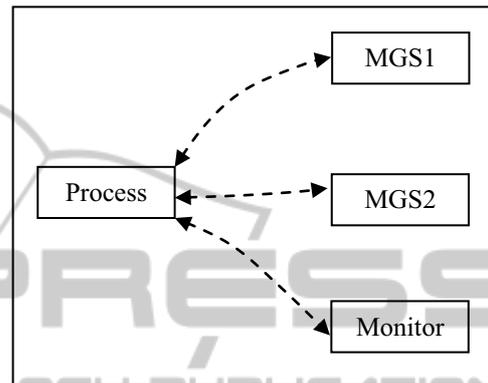


Figure 2: A system model representing high granularity service (HGS).

Table 2 describes the two additional types of compound services found at the medium and high granularity level of large Web services system i.e. i.e. MGS and HGS.

Table 2: Types of compound services found in large Web services systems.

Service	Description
MGS	Medium granularity service.
HGS	High granularity service

To the best of our knowledge, most existing business process measures are actually model-level (i.e. target only models of atomic business processes). Since large enterprise-wide and cross-enterprise Web service systems present a new measurement problem, we propose some size measures for such systems in the next section.

4 PROPOSED MEASURES

The following sections present the proposed size measures. The measures are based on common intuition that combining two or more components results in a larger component. This approach is also supported in literature by several authors such as Briand et al. (1996) who discusses on module

additivity. Our methodology, therefore, is to identify a suitable size measure for low-level granularity measurement of business processes. After measuring all the services interacting in a system, then we obtain their summation as a measurement result for the next level system.

As an example, we use the NOBA measure to calculate the size of a service by simply counting the number of basic activities in it. Thus, the size of the Figure 1(a) is 5 since the process has got five activities in it i.e. NOBA (Process) = 5. Similarly, the size of the Figure 1(b) is 5 since the request/reply service has got 5 basic activities in it i.e. NOBA (request/reply service) = 5.

4.1 Number of Basic Activities in System (NOBAS)

The medium granularity service (MGS) is a system composed of all services interacting together to achieve a common goal. To evaluate MGS, we obtain a summation of the number of basic activities of all services in the system using the formula shown in Eq. 1.

$$NOBAS(MGS) = \sum_{s=1}^n NOBA(s) \quad (1)$$

Where s is a service of any of the types described in Table 1 and n is the number of services in the system.

For example, we calculate the number of basic activities in the MGS system in Figure 1 as follows:

$$NOBAS = NOBA(\text{process}) + NOBA(\text{request/reply}) = 5 + 5 = 10.$$

HGS systems are extensions of MGS systems. In a HGS, all MGSs are treated as regular services and invoked alongside simple atomic services. Consequently, the same formula for MGS can apply as shown in Eq. 2.

$$NOBAS(HGS) = \sum_{s=1}^n NOBA(s) \quad (2)$$

Where s is a service of any of the types described in Table 1 and Table 2 and n is the number of atomic services in the system.

4.2 Number of Control-flows in System (NOCS)

The number of control-flows in the system is an extension of the number of structured activities in a business process (NOSA) (Muketha et al., 2010).

While this existing metric factored only those control-flows that are present in a business process model, we propose to count all control-flows in a system. Such a measure will be more useful to managers rather than designers of individual processes.

To count the number of control-flows in a system (either MGS or HGS), we simply obtain a summation of the control-flows in all atomic services in the system as shown in Eq. 3.

$$NOCS = \sum_{s=1}^n NOSA(s) \quad (3)$$

Where s is a service of any of the types described in Table 2 and n is the number of services in the system.

As an example, number of control-flows in the system in Figure 1 may be computed as follows:

$$NOCS = NOSA(\text{process}) + NOSA(\text{request/reply service}) = 1 + 2 = 3.$$

4.3 Number of Invocations in System (NOIS)

The number of invocations in the system is a count of the total number of times the client and/or the process invokes its partners. It also involves cases where services invoke other services before replying to the client is involved. Invocations introduce more complexity to the system that other interactions such as replying.

To count the number of invocations in the system, we use the formula in Eq. 4.

$$NOIS = |Inv| \quad (4)$$

Where the length of Inv is the set of all invocation links in the system.

As an example, the number of invocations in the system in Figure 1 is 2 i.e. client request and the request from process to the request/reply service.

4.4 Total System Size (TSS)

The TSS is computed as the summation number of basic activities in system, the number of control-flows in the system and the number of invocations in the system. Invocations are considered in the size they are a type of control-flow. Thus, this measure seeks to capture the sum of all elements in the system as well as their interactions as shown in Eq. 5.

$$TSS = NOBAS + NOCS + NOIS \quad (5)$$

As an example, the TSS of the system in Figure 1 may be computed as follows:

$$TSS = 10+3 + 2 = 15.$$

4.5 System Control-flow Density (SCD)

The density of control-flows in a system has been linked to programmer style, where for instance, one programmer may write a long program with few control-flows while another may have more. For this reason, the overall size as indicated by NOBAS is not necessarily indicative of complexity. SCD is a measure of the degree of control-flows in relation to the total system size. We calculate the SCD of a system as shown in Eq. 6.

$$SCD = NOCS / NOBAS \quad (6)$$

As an example, the SID in the system in Figure 1 is computed as follows:

$$SCD = (1+2)/10 = 0.3.$$

4.6 System Invocation Density (SID)

It is important to know the relative number of invocations in the system in addition to knowing the total invocations. SID is a measure of the degree of invocations in relation to the total system size. We calculate the SID of a system as shown in Eq. 7.

$$SID = NOIS / NOBAS \quad (7)$$

As an example, the SID in the system in Figure 1 is computed as follows:

$$SID = 2/10 = 0.2.$$

5 RESULTS

All newly defined measures need to be validated both theoretically and empirically. The size measures proposed in this paper has been validated theoretically based on Briand's framework (Briand et al., 1996), and the results are presented in this section.

Briand's framework proposed five metrics validation categorizes, namely, size, length, complexity and coupling and cohesion (Briand et al., 1996). Since our measure is a size measure, we used the three size properties in Briand's framework to validate it.

Size 1: Non-negativity. The size of a Web service system cannot be negative, but can be null if

the system has got no services in it i.e. NOBAS (HGS) ≥ 0 . Similarly, NOCS (HGS) ≥ 0 , and NOIS (HGS) ≥ 0 , and TSS (HGS) ≥ 0 . In addition, the density measures are evaluated under size properties since they are derived from size measures. Thus, SCD (HGS) ≥ 0 and SID (HGS) ≥ 0 since non negative but possibly null values of their base measures leads to a value of zero.

Size 2: Null value. The size of a Web service system is null if system is empty i.e. if it has got no service nodes in it, then NOBAS (HGS) = 0. Similarly, NOCS (HGS) = 0, and NOIS (HGS) = 0, and TSS (HGS) = 0, and SCD (HGS) = 0, and SID (HGS) ≥ 0 .

Size 3: Module additivity. The size of a Web service system is equal to the sum of the sizes of two of its modules. For example in Figure 2 the size of the HGS system is equal to the sum of the sizes of the process, MGS1, MGS2, and monitor services. Clearly, the proposed measures satisfy this property since measurement is based on summations of atomic services within the system.

The fact that our metrics satisfy all three size properties as proposed in Briand's measurement framework is an indicator of sound structural definition of the proposed measures. Table 3 presents a summary of these results.

Table 3: Summary of Theoretical Results.

Measure	Size 1	Size 2	Size 3
NOBAS	Yes	Yes	Yes
NOCS	Yes	Yes	Yes
NOIS	Yes	Yes	Yes
TSS	Yes	Yes	Yes
SCD	Yes	Yes	Yes
SID	Yes	Yes	Yes

Yes = satisfied property

The proposed measures focus on orchestration-based Web service systems. Although a BPEL environment is assumed for all the examples given, the proposed measures could also apply to other orchestration-based programming environments.

Generally, large values of the measures should be taken as a pointer that the system being evaluated is risky and error-prone. We have not established a threshold for the measure, because extensive empirical studies are needed first before this can be possible. However, our approach to measure based on service granularity levels is a first step towards evaluating very large systems. For instance, systems that are in the high granularity level should be taken as being in the risk-level category.

6 CONCLUSIONS

In this paper, we have proposed size measures for large Web service systems. We have provided several examples in order to show how the measures might be calculated. We have also validated the measures theoretically using Briand's measurement framework. Theoretical results show that the proposed measure satisfied all three size properties from Briand's framework, which implies that it is a structurally sound measure.

Future work is to conduct empirical studies in order to analyze the value of the proposed measures in relation to external quality characteristics for Web services such as maintainability.

REFERENCES

- Briand L. C., Morasca S. and Basilli V. R. 1996 Property-Based Software Engineering Measurement. *IEEE Transactions on Software Engineering*, Vol. 22, pp.68-86.
- Cardoso J., 2008. Business Process Control-Flow Complexity: Metric, Evaluation, and Validation. *International Journal of Web Services Research*, Vol. 5, pp.49-76.
- Daniel F., and Pernici B. 2006. Insights into Web Service Orchestration and Choreography. *International Journal of E-Business Research*, Vol. 2, 2006, pp.58-77.
- Cardoso J., Mendling J., Neumann G., and Reijers H.A., 2006. A Discourse on Complexity of Process Models (Survey Paper), *LNCS*, Vol. 4103, pp.115-126.
- Charfi A., and Mezini M. 2004. Aspect-Oriented Web Services Composition with AO4BPEL", *LNCS*, Vol. 3250, pp.168-182.
- Gruhn V., and Laue R. 2006. Adopting the Complexity Measure for Business Process Models", *5th IEEE International Conference on Cognitive Informatics*, pp.236-241.
- Mendling J., and Neumann G., 2007 Error Metrics for Business Process Models. *19th International Conference on Advanced Information Systems Engineering (CAISE'07)*, pp.53-56.
- Michelson B., 2005. Business Process Execution Language (BPEL) Primer: Understanding an Important Component of SOA and Integration Strategies. <http://www.psgroup.com>
- Modafferi S., and Conforti E. 2006. Methods for Enabling Recovery Actions in WS-BPEL. *LNCS*. Vol. 4275, pp.219-236.
- Muketha G. M., Ghani A.A.A., Selamat M.H. and Atan R., 2010. Complexity Metrics for Executable Business Processes", *Information Technology Journal*, Vol. 9, pp.1317-1326.
- Munoz L., Mazon J., and Trujillo J. 2010. A Family of Experiments to Validate Measures for UML Activity Diagrams of ETL Processes in Data Warehouses. *Information and Software Technology*, Vol. 52, pp.1188-1203.
- Rolon E., Cardoso J., Garcia F., Ruiz F., and Piattini M. 2008. Analysis and Validation of Control-Flow Complexity Measures with BPMN Process Models. *LNBP*, Vol. 29, pp.58-70.
- Rud D., Schmietendorf A., and Dumke R., 2006. Product Metrics for Service-Oriented Infrastructures, In A. Abran, M. Bundschuh, G. Büren, and R. Dumke, eds., *Applied Software Measurement. Proc. of the International Workshop on Software Metrics and DASMA Software MetrikKongress (IWSM/MetriKon 2006)*. Magdeburger Schriften zum Empirischen Software Engineering, Potsdam, Germany, Hasso-Plattner-Institut, Shaker Verlag, November, pp.161-174.
- Zhang Q., and Li X., 2009. Complexity Metrics for Service Oriented Systems. *Second International Symposium on Knowledge Acquisition and Modeling*, *IEEE*, pp.375-378.
- Zheng Y., Zhou J., and Krause P. 2007. Analysis of BPEL Data Dependencies. *33rd EUROMICRO Conference on Software Engineering and Advanced Applications*. *IEEE*. pp.351-358.