

Building Service Composition based on Statistics of the Services Use

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Abstract. A distributed computing environment is being formed within the geoportals IDSTU SB RAS, which is mainly used for processing spatial data. Processing services that implement the WPS standard, and data storage services, a catalog of WPS services, and service execution systems and their compositions have been developed. In the system for executing services statistics data is collected on user-made service calls. The analysis of the obtained statistical data is carried out in order to identify the data transfer between services calls and build service compositions.

Keywords: SOA, WPS, spatial data processing, semantic network, composition of services.

1 Introduction

Now the number of Web services [1] is actively growing, providing processing and publication of data. For example, remote sensing data services, spatial data processing and publishing services. Storage and publication services are creating. In particular, these are cloud data storages, such as Internet services: Dropbox, OneDrive, Google Drive, iCloud, Yandex.Disk, Cloud Mail.Ru, etc. The developed services greatly simplify applying various algorithms and data acquisition, excluding the often complicated process of installing and configuring software. Standards and specifications have been developed and applied that unify the call of services, the transmission and reception of data, for example, the OGC WPS spatial data processing standard [2], the set of REST specifications [3]. Service catalogs are created that contain meta-information and allow you to search for necessary services. The developed services are actively used by users to solve many problems.

Often, the solution of a problem may require repeating a certain sequence of services calls, where the user enters the same parameter values for each service. To automate the user's work, it is necessary to create a composition of Web services [4]. The standards for setting service compositions are BPEL (Business Process Execution Language), BPMN (Business Process Modeling Notation), BPML (Business Process Management Language), and XPDL (XML Process Definition Language). In addition,

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tion, programming languages are actively used to define service compositions. There are a number of workflow systems that allow you to create service compositions and have achieved significant results in various subject areas: Pegasus [5], Kepler [6], Swift [7], KNIME [8], Taverna [9], Galaxy [10], Trident [11] and Triana [12], Everest (Mathcloud) [13], CLAVIRE [14]. For processing spatial data, Geo-Processing Workflows is actively used [15]. The generally accepted standard for determining the composition of services is Directed acyclic graph (DAG) [16, 17], in which the vertices are the service calls, and the arcs are the dependencies between services according to the data. Creating a service composition is a complex process that requires programming skills from the user. The authors propose a method that automates the creation of service compositions based on statistical data of the services use by the user. This method allows you to connect single service calls to each other, to determine the dependencies from the data.

2 Statistics collection

Within the framework of the geoportal IDSTU SB RAS, a distributed computing environment is being formed, which is mainly used for processing spatial data. Processing services that implement the WPS standard, and data storage services, a catalog of WPS services, and service execution systems have been developed. The interaction between processing services and data storage services is implemented, which allows simplifying the use of services by the user. In the system for executing services and their compositions, statistics of service calls is collected. The statistics data includes the name of the service and its address, values of input and output parameters, service execution time, successful execution, execution errors, etc.

3 Definitions

We introduce the following notation necessary to describe the operation of the method.

$s = \langle \text{name}, I, O \rangle$ is the service, where $\langle \text{name} \rangle$ is the name of the service, I is the set of input parameters, O is the set of output parameters of the service. Below we will denote $s.I$ and $s.O$ parameters that belong to a particular service. A user service call has input parameter values specified by the user and output parameter values.

$t = \langle V_I, V_O \rangle$ - service s call with values V_I, V_O of parameters $s.I$ and $s.O$.

Log is the set of successful service calls made by users. The cause of the unsuccessful service call may be incorrect parameter values. Therefore, statistics are filtered and only successful calls are left. Based on the Log data, it is necessary to form a set of service compositions in the form of DAG.

4 Defining data transfer between two service calls

To recognize a service composition, it is necessary to determine the presence of data transfer between two service calls. The determination of the relationship between two service calls can be made based on the analysis of parameter values. Parameters of service call can be divided into input and output. It is assumed that if the value of the output parameter matches the value of the input parameter, then there may be data communication between service calls. Parameter values can be string, numerical type, or resource URIs. Usually these are URLs to files, data services, etc. Each URI used to receive data uniquely identifies the transmitted data. Accordingly, using URI parameters, data transferring from one service call to another can be determined. Analysis of string and numerical parameters is more difficult, because of matching the values may be casual. Currently, string and numeric parameters are not considered.

Here is an algorithm for checking data transfer between two service calls.

Input: t_i, t_j - two service calls

Output: flag - connection between service calls and by what parameters

```
function islinked
  flag <- False
  for each vm in  $t_i.VO$ :
    for each vn in  $t_j.VI$ :
      If (typeof vm = file and typeof vn = file and vm = vn) then
        flag <- True
        vn.linkedwith <- vm
      end if
    end for
  end for
  return flag
```

This algorithm compares the output parameters of one service call with the input parameters of another service call. If the URL values of the parameters match, the connection between service calls is established. In addition, it is determined by what parameters they are connected.

5 Building service compositions based on statistics

To analyze the set of successful calls to Log services, the following algorithm is used, at the output of which we obtain a general directed acyclic graph DAG_{gen} , where calls to t_i services are connected by data. Additionally, we obtain the values of the parameters of services entered by the user.

Input: Log

Output: list - semantic relationships between services

for each t_i **in** Log:

for each t_j **in** Log:

```

If islinked (ti, tj) then
    add (ti, tj) into list
end for
end for

```

A user can perform a data-connected sequence of service calls that forms a connected subgraph D_i (graph connectivity components) in DAG_{gen} that is not connected to other subgraphs. Each such subgraph D_i is a special case of the implementation of the composition of services. To search for D_i subgraphs, the width search algorithm is used. Moreover, the complexity of the search is linear. It depends from the sum of the number of vertices and the number of edges of the graph DAG_{gen} .

The user can often repeat the same sequence of service calls. This leads to the creation of several D_i subgraphs in which calls of the same services are made, but the parameter values differ, i.e. creating isomorphic subgraphs D_i , considering only services and data transfer. Checking for isomorphism of graphs is carried out by traversing oriented subgraphs.

As a result of the execution of the algorithm, we obtain several sets of isomorphic $IDAG_i$ subgraphs. The number of D_j subgraphs in $IDAG_i$ is an estimate of the frequency of use of this services composition. Next, filtering isomorphic $IDAG_i$ subgraphs is carried out, for which the number of vertices is more than one and there is at least one call to the service for publishing data. The resulting composition of services is shown in Fig. one.

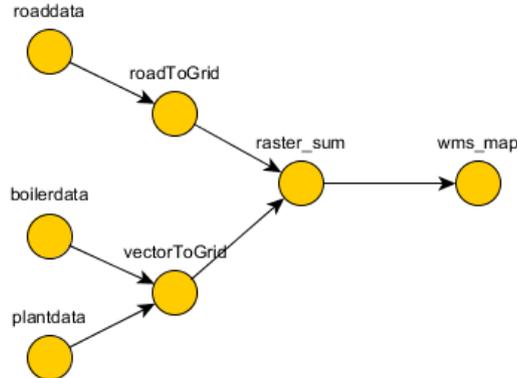


Fig. 1. The resulting services composition.

6 Conclusion

The proposed method, in contrast to existing approaches, uses statistics to build service compositions. To build a composition, the user only needs to sequentially perform the necessary services. The method automatically connects single service calls to each other, determines data dependencies. The application of the method is focused on the automation of frequently repeated user actions.

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