

Comprehensive quality - aware Automated Semantic Web Service Composition

Chen Wang, Hui Ma, and Aaron Chen

School of Engineering and Computer Science,
Victoria University of Wellington, New Zealand
{Chen.Wang, Hui.Ma, and Aaron.Chen}@ecs.vuw.ac.nz

Abstract. Semantic web service composition has been a prevailing research area in recent years. There are two major governance challenges faced by researchers. One is semantic matchmaking for discovering interoperable web services, which demands a rich machine understanding language such as OWL-S (Web Ontology language for Web Services), WSML (Web Service Modeling language, SAWSDL (Semantic Annotations for WSDL and XML Schema). This interaction of web services enables us to pair service functionalities through understanding and reasoning those semantic descriptions. Optimisation problems are another challenge. Many scholars have looked into nonfunctional optimisation problems in QoS-aware web service composition applying AI planning and Evolution Computing techniques. However, it is not sufficient without considering good balance between functional and nonfunctional quality in most scenarios. Therefore, the consideration in these two quality dimensions is a must, so that service composition needs to deal with optimisation problems in both quality of semantic matchmaking and QoS. This paper develops a more applicable way of measuring a proposed comprehensive quality model in both semantic matchmaking and QoS for automated semantic web service composition.

1 Introduction

Service-oriented computing (SOC) is a novel computing paradigm that employs services as fundamental elements to achieve agile development of cost-efficient and integratable enterprise applications in heterogeneous environments [20]. Service Oriented Architecture (SOA) is capable of abstractly realising service-oriented paradigm of computing. Since its inception, SOA has been contributing to the reuse of software components, from functions to units, and from units to services during the development in SOA [4]. One of the most typical realisations of SOA is *web service*, which is designated as “modular, self-describing, self-contained applications that are available on the Internet” [5]. Several standards play a significant role in registering, enquiring and grounding web services including particularly UDDI, WSDL and SOAP [8].

Web service composition pertains to a combination of multiple web services to provide a value-added composite service that accommodates customers’ ar-

bitrarily complex requirements. This application is developed by integrating interoperable and collaborative functionalities over heterogeneous systems. Due to the increase in the number of large-scale enterprise applications, the number of Web services has increased substantially and unprecedentedly. Therefore, manual and semi-automated web service composition are considered to be less efficient while automated web service composition enjoys less human intervention, less time consumption, and high productivity.

Two most notable challenges for web service composition are (1) ensuring interoperability of services and (2) achieving QoS optimisation [8]. *Interoperability* of web services presents challenge (1) in syntactic and semantics dimensions [8]. The syntactic dimension is covered by the XML-based technologies (such as *WSDL*, *SOAP*). The semantics aspect, on the other hand, demands further research. Through ontology-based semantics [18], web services can understand and better collaborate with each other. Historically, there are many ontologies languages and formats for semantic service descriptions, such as *OWL-S* [15], *WSML* [9], and *SAWSDL* [12]. The logical characteristics in reasoning makes “machine understanding” possible through identifying and matching semantic similarity in input/output parameters of web services in heterogeneous environments. The second challenge (2) is related to finding *optimised solutions* to Quality of Service (*QoS*) over composite web services. This problem gives birth to *QoS-aware service composition* that considers the composition of service-level agreements [21] involving a collection of SLA rules and policies for supporting QoS-based composition.

Existing works on service composition focus mainly on addressing the two challenges above. A lot of works have been conducted to optimise the quality of compositions under a pre-defined abstract workflow, which is generally considered as *semi-automated Web service composition* approach. Meanwhile, many research works consider the possibility of generating a composite plan automatically in discovering and selecting suitable web services, which are deemed to NP-hard [17]. *Semantic web services composition* is distinguished from the syntactic service composition, with the hope of eliminating conflicts by the semantic level of web services. In the past few years, substantial works have been done on semantic web service composition [8,13]. However, few works have enabled truly automatic semantic web service composition, where both QoS and quality of semantic match making will be optimised simultaneously. However, customers often prefer a service composition that best matches their requirement with highest QoS.

The overall goal of this paper is to *develop an comprehensive quality-aware automated semantic web service composition that satisfactorily optimises both functional and non-functional requirements*. Particularly, this paper considers the quality of semantic matchmaking quality and QoS and two structural constructs (sequence and parallel) for building optimised composition solution using Particle Swarm Optimisation (PSO). We aim to provide a more applicable way to measure semantic matchmaking in automated semantic web service composition. We will achieve three objectives in this work as follows:

1. To propose a comprehensive quality model that address QoS and semantic matchmaking quality in considering different matching types with corresponding semantic similarity. This approach is consider to be a more applicable and effective way of finding an optimised solution in semantic web service composition.
2. To propose a PSO-based service composition algorithm that utilises the proposed quality model. In PSO, weighted graphs with edges and vertices associated with quality of semantic matchmaking and QoS respectively represent the solutions of semantic web service composition in the searching space, and the weighted graph solutions are decoded from an optimised service queue with the consideration of the proposed comprehensive quality.
3. To evaluate the performance of comprehensive quality awareness semantic web service composition by utilising benchmark datasets from Web Services Challenge 2009 (WSC09) [11].

The remains of this paper are organised as the following: Sect. 2 provides related works in the semantic web service composition; Sect. 4 presents an overview of comprehensive quality-aware semantic automated web service composition consisting of comprehensive quality evaluation model, proposed weighted graphs as solution presentation and PSO-based service composition algorithm; Sect. 5 describes the experiments conducted to test the effectiveness of proposed model; Sect. 6 presents and analyses the experiment; Sect. 7 concludes our work.

2 Related Work

Semantic web service matchmaking. Substantial work [2,16,25,26,28] on web service composition focused primarily on non-functional requirements consistently neglecting functional requirements. However few researchers addressed both functional and nonfunctional requirements at the same time in web service composition. To the best of our knowledge, [6,13] reported some recent attempts over published. Semantic matchmaking utilises Description Logic(*DL*) [1] reasoning between input and output parameter concepts of web services to ensure a semantic matchmaking. It is considered to be a process consisting of seeking similarity of parameters (i.e., input and output) of web services and mapping between two knowledge representations encoded utilising the ontology.

In [22], three approaches have been studied regarding the similarity measures using taxonomies: (1) The first one is based on nodes, in which similarity is determined by the information content of the nodes; (2) the second is entirely based on edge, where concept distance in a hierarchy structure is evaluated, and (3) the hybrid approach features a combination of (1) and (2). A new similarity measure based on edge counting in taxonomy is introduced in [22], which extends similarity measure defined by Wu and Palmer [27]. Neighbourhood concepts are considered in their model when $\lambda = 1$ (default value as 0).

In [13], the quality of matchmaking problem is transferred to measure the quality of semantic links $sl_{i,j} \doteq \langle s_i, Sim_T(Out_{s_i}, In_{s_j}), s_j \rangle$, one possible measure is applied to the degree of similarity using Common Description rate of a

semantic link, where Extra Description $In_{s_x} \setminus Out_{s_y}$ and Least common subsume $lcs(Out_{s_i}, In_{s_j})$ are required to be pre-calculated. They also chained the web service with five well-known matching types: Exact, Plug-in, Subsume, Intersection, and Disjoint. Therefore, the quality of the semantic link is estimated as quality criteria q_m associated with their corresponding quality of vector q_{cd} . However, the weakness of semantic link quality is that calculating Extra description and Least common subsume requires well and completely defined ontology in class, class axioms and properties in OWL2. This makes it difficult to measure semantic matchmaking quality in most available semantic web services, as it takes huge cost and time for the domain experts to establish required ontology. In comparison, we introduce a more applicable comprehensive quality model in a fully automated approach.

The paper [6] covers the concerns in both functional and non-functional requirements in design time of semantic-based automatic service composition, where a GA-based approach with four unusual independent fitness functions are designed to solve the problems with these concerns. In details, a sequence of fitness functions are used in the binary selection of chromosomes, rather than a single objective consisting of both functional and non-functional quality. Apart from that, match types are ignored in their quality evaluation of semantic matching.

QoS-aware EC approaches. Evolution Computing techniques are widely used to solve optimisation problems. Genetic Programming (GP) [25,26] is a typical EC technique for automated web service composition. This GP-based approach utilises tree representations, on which service and constructs are represented as terminal nodes and functional nodes respectively. The crossover and mutation operations reproduce various individuals while ensuring the correctness of structure. In [28], the author optimises the overall quality of service composition by using a fitness function, which is also liable for the correctness in functionality through penalising infeasible solutions.

To simplify the checking of constraints for solutions, an indirect PSO-based approach was introduced in [25]. A general graph is used as representation as composite solution in their PSO algorithm considering QoS optimisation. However, this solution does not consider distinguished matchmaking quality, which could lead to over-general outputs to be produced by selected services. In fact, customers' perspectives, application domains and ontology granularity all could have significant impact on the outputs requested by users. In some scenarios, the output is too broad to bear any specific meaning for the customers, even though those web services selected leads to a very good overall QoS. Therefore, we fill the gap by considering different matching types and parameter similarity to determine the overall quality of web service composition. Weighed graphs are also utilised in PSP to facilitate a new representation of web service composition adopted from [25].

3 Problem Description

The purpose of web service composition is to accomplish an arbitrarily complex task fulfilling customer's requirement, which could be denoted as a composite goal: $Comp.G(F(T_{Input}, T_{Output}), NF(T_{QoS}))$. This overall composite goal is demonstrated in two parts. The first functional part focus on transferring a given task input or input set to the desired Task output or output set. It typically refers to users' functional requirement. Another nonfunctional part specifies the acceptable level of composite quality of service. To accomplish the composite goal, two stages are involved: services discovery and service selection. Firstly, service discovery is to find matched web service: $S_n(F(S_{Input}, S_{Output}), NF(S_{QoS}))$ from a Service Repository: $S = \{S_1, S_2, \dots, S_n\}$ with the given T_{Input} . If no atomic web service could satisfy the composite goal, a combination of web services will be found to meet $Comp.G$. To ensure the composite solution returns the desired $Comp.G$, we consider a comprehensive quality model for service discovery and selection, where challenges (1) and (2) mentioned in Sect. 1 are also addressed in Sect. 4.

3.1 Semantic Web Service matchmaking Type

The semantic service matchmaking aims to discover appropriate services from service repository in view of customers' functional requests. A semantic web service is defined by $S(F(S_{Input} \in C_1, S_{Output} \in C_2), NF(S_{QoS}))$, where both Input and Output are linked to concept C_1 and C_2 in an ontology (O) respectively, satisfying $O = \{C, Taxonomy\}$. A web service matching process is to match the output and input concepts of two services according to the Taxonomy within an Ontology (O). To measure the quality of semantic matchmaking, different matching levels are typically considered in the literature [19]. To understand these levels, let us define two web services associated with concept-related parameters in a particular domain. $S_1 (F(S_{Input} \in C_1, S_{Outputs} \in C_2), NF(S_{QoS}))$ and $S_2 (F(S_{Input} \in C_3, S_{Output} \in C_4), NF(S_{QoS}))$ and an Ontology(O) with C_1, C_2, C_3 , and C_4 . The matching levels to be considered are:

- *Exact* (\equiv): Output of Web service S_1 and Input of Web service S_2 are Exact match ($S_{output} \in S_1 \equiv S_{input} S_2$), if Concept C_2 and Concept C_3 are equivalent.
- *Plugin* (\sqsubseteq_n): Output of Web service S_1 and Input of Web service S_2 are Plugin match ($S_{output} \in S_1 \sqsubseteq_n S_{input} \in S_2$), if Concept C_2 is a sub-concept of Concept C_3 , and $n = \{1, 2, \dots, n\}$ presents the levels of children concepts ($n = 1$ stands for direct children).
- *Subsume* (\supseteq_n): Output of Web service S_1 and Input of Web service S_2 are Subsume matched ($S_{output} \in S_1 \supseteq_n S_{input} \in S_2$), if Concept C_2 is a sub-concept of Concept C_3 , and $n = \{1, 2, \dots, n\}$ presents the levels of parent concepts ($n = 1$ stands for direct parent).
- *Fail* (\perp). Output of Web service S_1 and Input of Web service S_2 are not matched (Fail) ($S_{output} \in S_1 \perp S_{input} \in S_2$), if none of the previous matches discovered.

3.2 Quality of Service and Composition Constructs

Currently, most of the optimisation problems [7,10,14,24] in web service composition are focusing on QoS, which covers aspects in non-functional requirements. This problem has been explored in both single objective and multi-objectives optimisation problems. Customers prefer lowest execution cost with highest response time and reliability that could be optimised simultaneously. According to [29], four most often considered QoS parameters are as follows:

- *Response time* (T) measures the expected delay in seconds between the moment when a request is sent and the moment when the results are received.
- *Cost* (C) is the amount of money that a service requester has to pay for executing the web service
- *Reliability* (R) is the probability that a request is correctly responded within the maximum expected time frame.
- *Availability* (A) is the probability that a web service is accessible.

The aggregation value of QoS attributes for web services composition varies with respect to different constructs, which reflects how services associated with each other in a service composition [29]. Here we consider two composite constructs: sequence and parallel constructs, in building the composite web service. The QoS calculation models are described as follows:

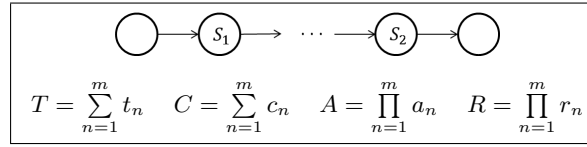


Fig. 1. Sequence construct and calculation of its QoS properties [28].

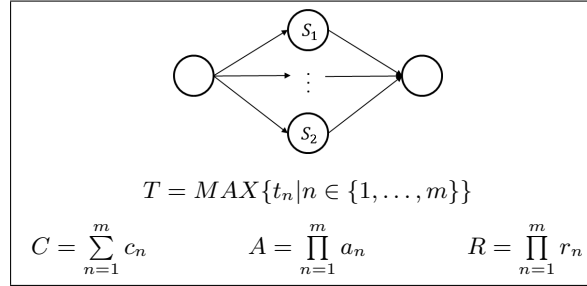


Fig. 2. Parallel construct and calculation of its QoS properties [28].

Sequence construct The composite web service executes each atomic service associated with a sequence construct in a definite sequence order. The aggregation value for total time (T) and total cost (C) is as the sum of time and cost

of web services involved respectively. The overall availability and reliability in a sequence construct are calculated by multiplying their corresponding availability and reliability of each web service in probability theory. This construct is shown in Fig. 1.

Parallel construct Web services in a parallel construct are executed concurrently. The QoS aggregation value for total cost, availability and reliability are the same as these in sequence construct while the Total time (T) is determined by the most time-consuming path in the composite flow of the solution. This construct is presented in Fig. 2.

4 Comprehensive quality-aware Semantic Automated Web Service Composition

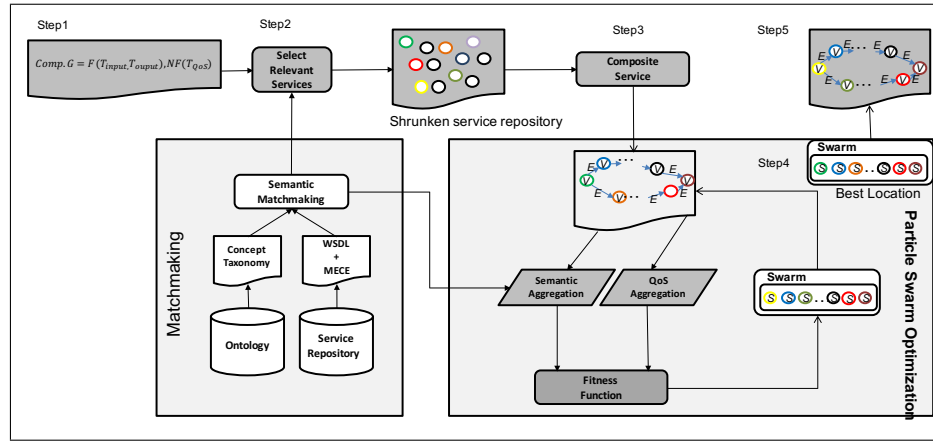


Fig. 3. Overview of the proposed approach.

In this paper, we propose a comprehensive quality model for automated semantic service composition, and optimise both the quality of semantic match-making and QoS. PSO has shown its efficiency in solving combinatorial optimisation problems. Therefore, we will employ a PSO-based approach, which is considered to be simple and efficient without penalising or repairing that often required by GP [24]. Fig. 3 shows the overview of our approach with five steps. Step 1: The composite process is triggered by a composite goal defined in Subsection 3, which describes customers requirements both functional and non-functional. Step 2: This composition goal are used to discover all relevant web services, which lead to a shrunk service repository that is subsequently used by PSO as a searching space. Step 3: A weighted graph representation is

randomly built up from the an initial service queue that mapped to the particle's location, interleaving with semantic matchmaking process. In the weighted graph, graph edges are assigned with semantic matchmaking quality as weights. Step 4: The fitness value of the weighted graph is evaluated to update the position of particle under PSO algorithm in Sect. 4.5, where the position is mapped to the index of service queue. later on, the updated service queue is used to decode a new weighted graph as the composition solution. Step 4. Lastly, the best position found in the searching space is selected and decode into the final optimised solution. This PSO-based approach is different from [25], as we use weighted graphs as solution presentation.

4.1 Semantic Matchmaking

To perform semantic matchmaking, we transfer a function match between $S_1 : S_{output} \in C_1$ and $S_2 : S_{output} \in S_b$ to a pair of concept match demonstrated in Sect. 3.1. The matching process attempts to determine semantic matching between the source concepts of C_1 and the target concepts of C_2 . Meanwhile, the quality of matched concepts are calculated in the quality model in subSection 4.2. The purpose of semantic matchmaking is to find more component services that could also potentially satisfy the quality of QoS with good functional quality.

The semantic matchmaking is achieved by utilising OWL2 and OWL-S or other semantic markup languages for web services. In this paper, we use MECE (Mediation Contract Extension) [3] and OWL-DL. MECE is considered to be an alternative semantic annotation for WSDL. MECE defines the service-related inputs and outputs with parameter-related concepts. OWL-DL is a sublanguage of OWL extended from RDF. It specifies semantic information of concepts involved in MECE.

4.2 Comprehensive Quality Model and Aggregation Matrix

In this paper, we propose a comprehensive quality model to evaluate the overall quality of semantic web service composition. This model overcome the disadvantages of current prevailing QoS-aware optimisation [2,16,25,26,28] that ignores quality of semantic matchmaking.

Semantic matchmaking model. Due to the discretisational characteristics of different match types and values assigned to matching types that driven by the cost of data integration and manipulation [13], partial ordering match types are considered to be one factor for the semantic matchmaking quality. For example, Exact matching type demands less time for computation compared to that of Plugin match. Another factor in our proposed model is concept similarity, which could be evaluated based on the edge counting method defined in [22]. This formula (1) is used to estimate the similarity between parameter-related output concept and parameter-related input concept for selecting web services. Therefore, given the quality match type and the similarity of two parameters-related concepts, the semantic matchmaking quality of matched parameters is

defined by Formula (2), where the value of $q(p_{mt})$ follows the same settings in [13]. Particularly, 1 (Exact), 0.75 (Plugin), 0.5 (Subsume) or 0.25 (Intersection).

$$q(p_s) = \frac{2N \cdot e^{-\lambda L/D}}{N_1 + N_2} \quad (1)$$

$$q(p_{sm}) \doteq (q(p_{mt}), q(p_s)) \quad (2)$$

Further more, edges represent services connections in our weighted graph representations, where assigned weight value $q(e_{sm})$ is considered to be semantic matching quality on edge level according to parameter aggregations. The weight value $q(e_{sm})$ is defined in 3, where $q(e_{mt})$ and $q(e_s)$ are the average value of concept-related parameters quality in $q(p_{mt})$ and $q(p_s)$ respectively.

$$q(e_{sm}) \doteq (q(e_{mt}), q(e_s)) \quad (3)$$

Comprehensive quality model. Compared to QoS evaluation model, the comprehensive quality model is established to investigate both functional and non-functional requirements. The comprehensive quality of our service composition representation refers to QoS of service vertices and Semantic matching quality of Edges in weighted graphs. Consequently, the comprehensive quality model is defined in Formula (4), which could be further broken down into Formula (5).

$$q_{cq} \doteq (q(e_{sm}), q(v_{QoS})) \quad (4)$$

$$q_{cq} \doteq (q(e_{mt}), q(e_s), q(v_a), q(v_r), q(v_c), q(v_t)) \quad (5)$$

Quality aggregate matrix. The quality aggregation is defined based on the constructs of composite web services, in consideration of functional and non-functional properties. The quality on construct level is further calculated by following the rules summarised in Table 1.

Table 1. Quality aggregate matrix for semantic web service composition

Composition Construct		Sequence	Parallel
QualityFactors	Functional	$Q(e_{mt})$	$\prod_{n=1}^m q(e_{mt})$
		$Q(e_s)$	$(\sum_{n=1}^m q(e_s))/m$
	NonFunctional	$Q(v_a)$	$\prod_{n=1}^m q(v_a)$
		$Q(v_r)$	$\prod_{n=1}^m q(v_r)$
		$Q(v_c)$	$\sum_{n=1}^m q(v_c)$
		$Q(v_t)$	$\max(q(v_t))$

4.3 Composition Weighted Graph

We defined our semantic web service composition solution as a weighted graph $WG = (V, E)$, where V is a set of services as vertex: $V = [S1, S2...Sn]$ and E

is a set of edges $E = e_1, e_2, \dots, e_n$. Each e is associated with $q(e_{sm})$ as weight value that mapped to a pair of quality values $q(e_{mt})$ and $q(e_s)$ on the edge level, and e_m is expressed as $(S_a, S_b) = q_{mt}, q_s$. Here we provide an example of the web service composition, which is described in Fig. 4. The data of composite web service flows from the start to the end, where five web services involved and linked with each other using edge connections. Besides that, $q(s_{mt})$ and $q(s_s)$ are calculated for all edges e_1, e_2, \dots, e_n assigned as weights.

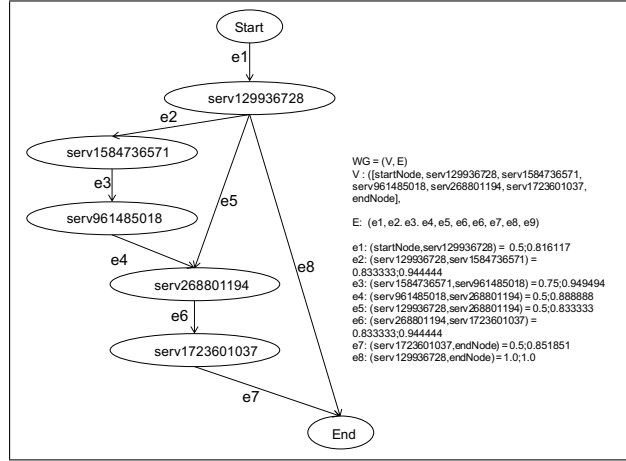


Fig. 4. Composition weighted graph solution

4.4 Fitness Calculation

The fitness value of a service composition solution is a weighted sum of all the quality related components in Formula (6), in which fitness value of 1 means the best comprehensive quality and 0 means the worst. For this purpose, MT , S , A , R , T , and C must be normalised so that the fitness value falls within the range from 0 to 1. Therefore, the service composition problem in this paper is treated as a fitness maximisation problem.

$$Fitness = w_1 \hat{MT} + w_2 \hat{S} + w_3 \hat{A} + w_4 \hat{R} + w_5 (1 - \hat{T}) + w_6 (1 - \hat{C}) \quad (6)$$

where $\sum_{i=1}^6 w_i = 1$

$$\hat{Q}_k = \begin{cases} \frac{Q_k - Q_{k,min}}{Q_{k,max} - Q_{k,min}} & \text{if } Q_{k,max} - Q_{k,min} \neq 0. \\ 1 & \text{otherwise.} \end{cases} \quad (7)$$

where $k = 1, 2, 3$, and 4 , where Q_1 as MT , Q_2 as S , Q_3 as A , and Q_4 as R .

$$\hat{Q}_j = \begin{cases} \frac{Q_{j,max} - Q_j}{Q_{j,max} - Q_{j,min}} & \text{if } Q_{j,max} - Q_{j,min} \neq 0. \\ 1 & \text{otherwise.} \end{cases} \quad (8)$$

where $j = 1$, and 2 , where Q_1 as T and Q_2 as C .

4.5 QoS-aware Semantic Web Service Composition Algorithm

ALGORITHM 1. Steps of the PSO-based Web service composition technique.

1. Map each relevant service to an index in the particle's position vector.
 2. Randomly initialise each particle in the swarm.
 - while** *max. iterations not met* **do**
 - forall** *particles in the swarm* **do**
 - 3. Create queue of services using the particle's position vector.
 - 4. Build a weighted graph using the queue.
 - 5. Calculate the fitness of the weighted graph.
 - if** *fitness value better than pBest* **then**
 - 6a. Assign current fitness as new pBest.
 - else**
 - 6b. Keep previous pBest.
 - 7. Assign best particle's pBest value to gBest, if better than gBest.
 - 8. Calculate the velocity of each particle according to the equation:
 - $v_{id} = v_{id} + c_1 * rand() * (p_{id} - x_{id}) + c_2 * rand() * (p_{gd} - x_{id})$
 - 9. Update the position of each particle according to the equation:
 - $x_{id} = x_{id} + v_{id}$
-

The overall algorithm investigated here is made up of a PSO-based web service composition algorithm 1 and a decoding algorithm 2, to generate a composition weighted graph. In algorithm 1, the idea is to translate the particle location produced by PSO into a service queue as an indirect representation of weighted graph, such that finding the best fitness of the weighted graph is to discover the optimised location of the particle in the search space. In PSO, the dimension of each particle is equals to the number of relevant web services. The index of each services is mapped to a separate location component in a particle. Services in a queue follow the ascending order, from which we decode a weighted graph using Algorithm 2. We select and connect service vertex to the start vertex from the service queue if the web service can satisfy the output of start vertex. Meanwhile, an Output Set is initialised and kept being updated using outputs from the newly added service vertex in the weighted graph, later on, more services vertices are connected with edges if services' inputs could be satisfied by any output in the Output Set. Finally, the end vertex is connected to the graph

ALGORITHM 2. Create a composition weighted graph from a queue

Input : $I, O, queue$
Output: composition weighted graph WG

1. Create *start* vertex with outputs O and *end* vertex with inputs I .
2. Create graph WG containing the *start* vertex.
3. Create available outputs set containing *start* outputs.
- while** *available outputs do not satisfy end inputs* **do**
 4. Get next candidate from queue.
 - if** *candidate inputs are satisfied by available outputs* **then**
 5. Create edge assigned with weight.
 6. Connect vertex to graph with the edge.
 7. Remove it from queue and go back to the queue's beginning.
8. Connect end vertex.
9. Remove dangling vertices from graph WG .
10. Remove dangling edges from graph WG .

return WG .

if the updated Output Set contains all end inputs. In addition, dangling service vertex and edges will be removed.

5 Experiment Design

In this section, a quantitative evaluation approach is adopted in our experiment design. The objectives of the evaluation are to (1) measure the effectiveness of the comprehensive quality model in automated semantic web service composition approach; (2) explore the impacts of the semantic matchmaking that contributes to overall composition quality; and (3) compare solutions generated by QoS-aware approach with our method.

We utilise benchmark dataset web service challenge 2009 (WSC09) [11] to perform the evaluation. WSC09 provides problems with five tasks corresponding to variable number of services, and ontologies. Therefore, it is a challenge dataset for measuring the scalability of our quality evaluation model. Table 2 presents the features of the WSC09 dataset. The number of concepts, individuals in the ontology and services in each data set is shown in the second, third, fourth column respectively. Also, we extend all the datasets with QoS attributes from service providers to enable our evaluation.

We run the experiment on computing grid comprising of almost 170 NetBSD (Unix operating system) workstations operated by the Sun Grid Engine. Experimentation is done using the approach explained in Section 4, and evaluated in the comprehensive quality model introduced in Subsection 4.2. The parameters were chosen based on general settings from [23] for our PSO-based approach, 30 particles evolved in 100 generations in the searching space, in which problem dimension's size equals to the number of relevant services in service repository. We run 30 times independently for each dataset. We configure weight of fitness function to properly balance functional side and nonfunctional side. Therefore,

Table 2. Features of the WSC09 datasets

Dataset	No.Concept	No.Individual	No.Service
WSC09 01	1578	3102	572
WSC09 02	12388	24815	4129
WSC09 03	18573	37316	8138
WSC09 04	18673	37324	8301
WSC09 05	31044	62132	15211

w_1 and w_2 are equals to 0.1 and 0.4, and w_3, w_4, w_5, w_6 are all set to 0.125 accordingly. However, the settings of the values does not impact the method. In general, weight settings are adjusted according to users' preferences.

6 Results and Analysis

6.1 Comparison Test

In this section, we analyse the composition solution generated by using our approach comparing with QoS-aware approach. In particularly, we show that our proposed approach can produce solutions that results better matchmaking, which meets user's goal better.

First, we look at mean value of Q_{mt} , Q_s and Q_{QoS} at optimum at the 100th generation for two approaches, shown in Table 3, where QoS-aware approach record Q_{mt} , Q_s while utilise fitness function $Fitness = w_1\hat{A} + w_2\hat{R} + w_3(1 - \hat{T}) + w_4(1 - \hat{C})$ when $\sum_{i=1}^4 w_i = 1$ without considering MT and S , and Q_{QoS} is normalised from 0 to 0.5 to make it comparable to Q_{QoS} in our approach. We observe an interesting pattern from the Table 3 using statistic analysis: mean value of Q_{mt} and Q_s at optimum by our approach is consistently higher than those by QoS-aware approach in Task 1, 3, 4, and 5 with an exception in Task 2 while the combination of mean Q_{mt} and Q_s at optimum in Task 2 is still considered to be higher in our approach with 0.1 and 0.4 weights respectively. Meanwhile, Q_{QoS} generated by QoS-aware approach can obtain a slightly higher value than that of our approach. In conclusion, we can perceive that our evaluation could find out better functional quality with a reasonable trade off in QoS.

Second, to compare the results generated from two evaluation approaches, we demonstrate an example solution that shows the differences in two composition weighted graphs representations under two different methods. The Fig. 5 shows two composition weighted graphs as the solutions to Task 3 with (1) QoS-aware approach and (2) Comprehensive quality-aware method respectively. Two approaches generate exactly the same service workflow structure while some service vertex and edges denoted in red are different. Therefore, we make a comparison of the quality among these different edges (e_1 to e_4) associated service vertex in terms of quality of semantic matchmaking and QoS attributes. We also look at ΔQ which reveals the amount of variation on quality between two methods,

Table 3. Mean Quality for comprehensive quality-aware methods and QoS-aware approach

WSC09		QoS-aware Evaluation	Comprehensive Quality Evaluation
Task1	Q_{mt}	0.189787 ± 0.039278686	$0.220594 \pm 0.009318 \uparrow$
	Q_s	0.884962 ± 0.014140	$0.893713 \pm 0.007769 \uparrow$
	Q_{QoS}	0.278730 ± 0.007786	0.279548 ± 0.007627
Task2	Q_{mt}	$0.001795 \pm 0.000719 \uparrow$	0.001135 ± 0.000633
	Q_s	0.906971 ± 0.005855	$0.927965 \pm 0.007306 \uparrow$
	Q_{QoS}	$0.239979 \pm 0.000578 \uparrow$	0.237845 ± 0.001439
Task3	Q_{mt}	0.158526 ± 0.014028	$0.245830 \pm 0.007761 \uparrow$
	Q_s	0.949109 ± 0.002331	$0.972765 \pm 0.002980 \uparrow$
	Q_{QoS}	$0.247002 \pm 0.000661 \uparrow$	0.245631 ± 0.000431
Task4	Q_{mt}	0.000000 ± 0.000000	$0.000001 \pm 0.000002 \uparrow$
	Q_s	0.880058 ± 0.006603	$0.930632 \pm 0.009043 \uparrow$
	Q_{QoS}	$0.242338 \pm 0.000516 \uparrow$	0.236677 ± 0.002211
Task5	Q_{mt}	0.000042 ± 0.000030	$0.000078 \pm 0.000031 \uparrow$
	Q_s	0.915933 ± 0.012888	$0.928189 \pm 0.002486 \uparrow$
	Q_{QoS}	$0.238189 \pm 0.000240 \uparrow$	0.237232 ± 0.000391

where the positive values means the benefits gained under our approach while the negative values means the trade-off. To demonstrate the benefits of the positive value, we analyse the smallest positive ΔQ corresponding to e_4 and demonstrate how the output and required input are different under two approaches in in Fig. 6. *Ser1640238160* and *Ser1947554374* are selected service vertex with output concept-related parameters *Inst795998200* and *Inst582785907* corresponding to *Con103314376* and *Con2037585750* respectively while *Inst658772240* are the required input related to *Con2113572083*, in which there exist *Con103314376* \sqsubseteq_2 *Con2113572083* and *Con2037585750* \sqsubseteq_3 *Con2113572083*. It is obviously that our approach select the service providing *Inst795998200* that are closer the users' requirement, which means our methods results in better semantic matchmaking quality.

6.2 Convergence Test

To analyse the effectiveness of our approach, we will study the convergence rate of the proposed method in this section to understand the convergence rate of five tasks in WSC09. To further study the overall effectiveness of evaluation model and explore impacts of all involved components in the fitness function. We analysis all the performances of these components in the whole evolutionary process in Fig. 7, in which the five tasks' experiment results are arranged in four groups consisting of average fitness, average matchType quality, average similarity quality and average QoS for generation 0-99 with optimum.

Firstly, the average fitness value with optimum is calculated by the average value of best fitness found in each generation over 30 independent runs. The value

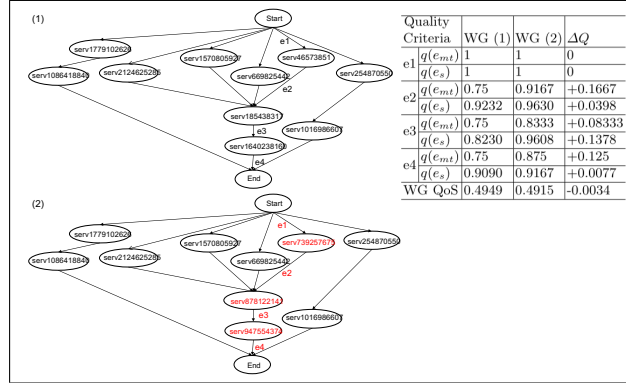


Fig. 5. Example Comparison of solutions to Task 3 under different approaches.

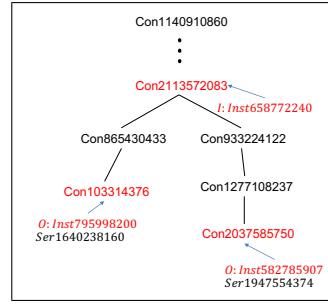


Fig. 6. Example Comparison of Edge 4 in Task 3 solution under two methods.

plotted in the first column group ranges from 0 to 1 in Fig. 7, and it is considered to be more optimised if its value is closer to value 1. We can see that there is a significant increase in the fitness values with optimum between generation 0 and generation 15-25, the remaining generation continues to produce a steady but moderate improvement in the fitness value and eventually reach a plateau with no further changes. The same behaviour is observed over the rest tasks. We also witnessed a fast convergence rate of the fitness value.

We also investigate the variation of quality of semantic matchmaking where average matchType quality with optimum and average similarity quality with optimum are studied from generation 0 to 99 in second and third column groups of Fig. 7. In these groups, both the average matchType quality with optimum and average similarity quality with optimum are the mean value corresponding to the best fitness value found in each generation over 30 independent runs. The dominant tendency of the marked value representing a similar characteristic compared to the average fitness value group trend in five tasks. However, there are some slight fluctuations over the generations.

Lastly, it is interesting to look at the average QoS per generation for the 30 independent runs when the functional quality heads up. Here average QoS with

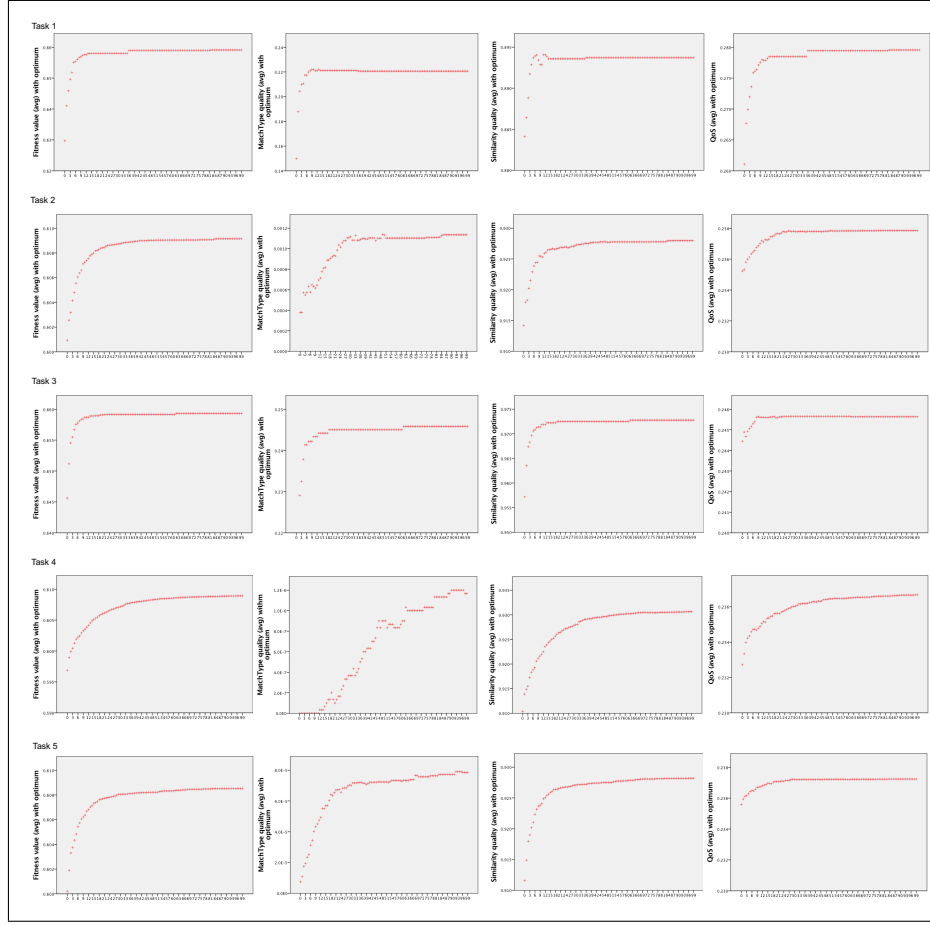


Fig. 7. Average Fitness, Average MatchType Quality, Average Similarity quality and Average QoS per generation with comprehensive quality optimum

optimum is also the value corresponding to the fitness value with optimum in each generation, and we display the results in the fourth column group in Fig. 7. It is obvious that the overall trend of QoS moves upward to a high constant level in five independent tasks. Additionally, we don't see too much trade-off from the QoS.

7 Conclusion

This work introduced a evaluation model for QoS-aware automated semantic automated web service composition that is based on a semantic matchmaking quality extends existing QoS. The results shows that our comprehensive quality model is proved to obtain better functional quality with a reasonable trade-off

in QoS. Future works in this area could explore the comprehensive quality with multi-objectives to maximising quality of matchmaking and to optimise QoS, and improve efficiency in calculating semantic quality.

References

1. Baader, F.: The description logic handbook: Theory, implementation and applications. Cambridge university press (2003)
2. Bansal, S., Bansal, A., Gupta, G., Blake, M.B.: Generalized semantic web service composition. *Service Oriented Computing and Applications* 10(2), 111–133 (2016)
3. Bleul, S., Comes, D., Kirchhoff, M., Zapf, M.: Self-integration of web services in bpm processes. In: *Proceedings of the Workshop Selbstorganisierende, Adaptive, Kontextsensitive verteilte Systeme (SAKS)* (2008)
4. Booth, D., Haas, H., McCabe, F., Newcomer, E., Champion, M., Ferris, C., Orchard, D.: Web services architecture. w3c working note. *W3C Working Notes* (2004)
5. Curbera, F., Nagy, W., Weerawarana, S.: Web services: Why and how. In: *Workshop on Object-Oriented Web Services-OOPSLA*. vol. 2001 (2001)
6. FanJiang, Y.Y., Syu, Y.: Semantic-based automatic service composition with functional and non-functional requirements in design time: A genetic algorithm approach. *Information and Software Technology* 56(3), 352–373 (2014)
7. Feng, Y., Ngan, L.D., Kanagasabai, R.: Dynamic service composition with service-dependent qos attributes. In: *Web Services (ICWS), 2013 IEEE 20th International Conference on*. pp. 10–17. IEEE (2013)
8. Fensel, D., Facca, F.M., Simperl, E., Toma, I.: *Semantic web services*. Springer Science & Business Media (2011)
9. Fensel, D., Lausen, H., Polleres, A., De Bruijn, J., Stollberg, M., Roman, D., Domingue, J.: *Enabling semantic web services: the web service modeling ontology*. Springer Science & Business Media (2006)
10. Huang, Z., Jiang, W., Hu, S., Liu, Z.: Effective pruning algorithm for qos-aware service composition. In: *2009 IEEE Conference on Commerce and Enterprise Computing*. pp. 519–522. IEEE (2009)
11. Kona, S., Bansal, A., Blake, M.B., Bleul, S., Weise, T.: Wsc-2009: a quality of service-oriented web services challenge. In: *2009 IEEE Conference on Commerce and Enterprise Computing*. pp. 487–490. IEEE (2009)
12. Lausen, H., Farrell, J.: Semantic annotations for wsdl and xml schema. *W3C recommendation, W3C* pp. 749–758 (2007)
13. Lécué, F.: Optimizing qos-aware semantic web service composition. In: *International Semantic Web Conference*. pp. 375–391. Springer (2009)
14. Ma, H., Wang, A., Zhang, M.: A hybrid approach using genetic programming and greedy search for qos-aware web service composition. In: *Transactions on Large-Scale Data-and Knowledge-Centered Systems XVIII*, pp. 180–205. Springer (2015)
15. Martin, D., Burstein, M., Hobbs, J., Lassila, O., McDermott, D., McIlraith, S., Narayanan, S., Paolucci, M., Parsia, B., Payne, T., et al.: *Owl-s: Semantic markup for web services*. W3C member submission 22, 2007–04 (2004)
16. Mier, P.R., Pedrinaci, C., Lama, M., Mucientes, M.: *An integrated semantic web service discovery and composition framework* (2015)
17. Moghaddam, M., Davis, J.G.: Service selection in web service composition: A comparative review of existing approaches. In: *Web Services Foundations*, pp. 321–346. Springer (2014)

18. O’Leary, D.: Review: Ontologies: A silver bullet for knowledge management and electronic commerce. *The Computer Journal* 48(4), 498–498 (2005)
19. Paolucci, M., Kawamura, T., Payne, T.R., Sycara, K.: Semantic matching of web services capabilities. In: *International Semantic Web Conference*. pp. 333–347. Springer (2002)
20. Papazoglou, M.P.: Service-oriented computing: Concepts, characteristics and directions. In: *Web Information Systems Engineering, 2003. WISE 2003. Proceedings of the Fourth International Conference on*. pp. 3–12. IEEE (2003)
21. Sahai, A., Machiraju, V., Sayal, M., Van Moorsel, A., Casati, F.: Automated sla monitoring for web services. In: *International Workshop on Distributed Systems: Operations and Management*. pp. 28–41. Springer (2002)
22. Shet, K., Acharya, U.D., et al.: A new similarity measure for taxonomy based on edge counting. *arXiv preprint arXiv:1211.4709* (2012)
23. Shi, Y., et al.: Particle swarm optimization: developments, applications and resources. In: *evolutionary computation, 2001. Proceedings of the 2001 Congress on*. vol. 1, pp. 81–86. IEEE (2001)
24. da Silva, A.S., Ma, H., Zhang, M.: A graph-based particle swarm optimisation approach to qos-aware web service composition and selection. In: *2014 IEEE Congress on Evolutionary Computation (CEC)*. pp. 3127–3134. IEEE (2014)
25. da Silva, A.S., Mei, Y., Ma, H., Zhang, M.: Particle swarm optimisation with sequence-like indirect representation for web service composition. In: *European Conference on Evolutionary Computation in Combinatorial Optimization*. pp. 202–218. Springer (2016)
26. da Silva, A., Ma, H., Zhang, M.: Graphevol: A graph evolution technique for web service composition. In: Chen, Q., Hameurlain, A., Toumani, F., Wagner, R., Decker, H. (eds.) *Database and Expert Systems Applications, Lecture Notes in Computer Science*, vol. 9262, pp. 134–142. Springer International Publishing (2015)
27. Wu, Z., Palmer, M.: Verbs semantics and lexical selection. In: *Proceedings of the 32nd annual meeting on Association for Computational Linguistics*. pp. 133–138. Association for Computational Linguistics (1994)
28. Yu, Y., Ma, H., Zhang, M.: An adaptive genetic programming approach to qos-aware web services composition. In: *2013 IEEE Congress on Evolutionary Computation*. pp. 1740–1747. IEEE (2013)
29. Zeng, L., Benatallah, B., Dumas, M., Kalagnanam, J., Sheng, Q.Z.: Quality driven web services composition. In: *Proceedings of the 12th international conference on World Wide Web*. pp. 411–421. ACM (2003)