

Semantics Service Composition Using Conceptual Graph for Addressing Imprecise Service Requirements

William Song

Computer Science

Durham University, UK

w.w.song@durham.ac.uk



Introduction

- Web Services has become an important research topic in the fields of the Service-Oriented Architecture (SOA).
- Automatic or semi-automatic service discovery, invocation and composition techniques are on demand.
- The Semantic Web Services seems to be the most promising way towards achieving automatic or semiautomatic service discovery, invocation, and composition.



Problems

- Insufficient usage context information: The current work are focusing on ontology based data type semantics and do not sufficiently address how a service is fitted into its usage context.
- Precise requirements required to locate services: In order to locate the required services, the current work requires precise service requirements which are difficult to be specified at the preliminary stage of the service discovery.
- Insufficient information about inter-relationship among service: The current work has not addressed the inter-relationships among services sufficiently, which makes the service discovery in an isolated manner.



We try to describe services by

- using the inter-service relationships.
- Using the contextual information.
- so easing service composition process.



Our proposal

- A Context-based Semantic Description Framework (CbSDF).
 - To describe services by the usage context aspect using Conceptual Graphs and Spider Model.
 - To use non-monotonic rules to describe the pre-conditions and effects of services and the conditions for service composition.
 - To search for services based on imprecise service requirements.



Example: Learning Resources

- A learning flow with learning resource specifications, but not the physical resources.
- Learning resources located dynamically at learning time based on the specifications.
- The diagram illustrates a learning course with learning resource specifications. The dotted lines represent the links dynamically established at learning time





Context-based Semantic Description Framework (CbSDF)

- The proposed CbSDF consists of four components:
 - Definitions of atomic and composite services
 - By having clear definitions of atomic service and composite service, we can identify what kind of information is relevant to describing a service
 - Service Conceptual Graphs
 - Give an overall and abstract description of the relationships between services and their related concepts.
 - Semantic Service Description Model (Spider Model)
 - Semantically describes service itself and the relations with other services.
 - Non-monotonic Rules
 - Describe the pre-conditions and effects of services and the conditions for service composition.



Conceptual Graphs

- A conceptual graph (CG) is a finite, connected, bipartite graph with nodes of one type called concepts and nodes of the other type called conceptual relations
 - The label of a concept node consists of two fields separated by a colon, [type: referent] i.e. [class: instance].
 - Conceptual Relations represent the relationships between concept nodes.
- Projection of CG
 - $\pi: v \rightarrow u$, where $\pi_u v$ is a sub-graph of *u* called a *projection* of *v* in *u*. π is called the projection operator. *v* describes a more generalised concept than *u*, $u \leq v$
 - Projection concept is important in CG matching and reasoning.

Dependency Graph of Learning

• A learning concept dependency graph G_d is a CG where the concept type is restricted to concepts within the Learning Object ontology

$$G_d = \langle C, R, \vec{E} \rangle, type(C) \in O$$

- C: a set of learning concept nodes; type(C) returns a set of leaf node concepts in the Learning Object ontology.
- R: a set of relation nodes that represent the relations among learning concept nodes, including pre-requisite relation type and conceptual relation type etc.
- $-\dot{E}$: a set of arcs that associate relation nodes with concept nodes.
- *O*: the Learning Object ontology.



Four Types of Semantics in Web^{niversity} Services

- Data Semantics
 - Formal definition of data in Input and output message.
- Functional Semantics
 - Formal definition of the web service capability.
- Non-functional Semantics
 - Formal definition of quantitative or non-quantitative constraints.
- Execution Semantics
 - Formal definition of execution flow of services of a process or of operations within a service.

Graphical Illustration of SSDM

- The notations used in the diagram are:
 - S_i=Service_i
 - Service' can be either the parent or ancestor of Service,
 - S₁, S₂, S₃, and S₄ are any services.
 - *I*, *P* is the Inputs and Pre-condition, and *O*, *E* is the Outputs and Effects.





Learning Objects Ontology

- The ontology represents the concepts of Learning Objects
- Based on the ACM/IEEE Computing Curriculum.
- Three levels: Area, Unit, and Topic.
- The leaf node of the ontology is a course or part of a course that can be directly taken by learners.





Non-monotonic Rules

- Reason for using non-monotonic rules
 - Handling unpredictable situation in a open service repository.
 - Exception handling.
- The non-monotonic rules are described using Defeasible Logic. A *defeasible theory DT* is a triple:

$$DT = (F, R, >)$$

- *F*: a set of facts;
- *R*: a finite set of rules;
- >: a superiority relation on *R*.
- The rules are divided into two categories:
 - General rules
 - Domain specific rules
- The rules are used in two ways:
 - Describe services pre-conditions and effects.
 - Validate service composition results: Trigger-able validation and Compose-able validation.



General Rules

- The general rules are normally used to construct and validate composite services and they are applicable to all the services, for example :
 - *r1*: if a service's pre-condition is satisfied, then normally it can be executed.

 $satisfy(S.preCon) \Rightarrow executable(S)$

- r2: if a service is not available, then definitely it cannot be executed.

 $\neg available(S) \rightarrow \neg executable(S)$

 r3: if two services are composed through input and output data flow, then normally the data types of the input and output are compatible, i.e. one is a same or sub-type of the other.

- The P_2^{n} Raschigher p_1^{n} p_1^{n} p_1^{n} p_2^{n} p_2^{n}



Domain Specific Rules

- The domain specific rules are normally used to describe the pre-conditions and effects of services and can only be applied to a specific domain, for example:
 - r1: if the service is supplied with a valid postcode, then normally the correct result will be returned. valid (postcode) \Rightarrow result(S)
 - *r2*: if the requested address is in UK, then this service is definitely applicable.

location $(UK) \rightarrow applicable(S)$



Two-Step Service Discovery Mechanism

- The first step is preliminary service discovery step using the CG matching technique.
 - Requirement \rightarrow CG
 - Match with Service Conceptual Graphs
- The second step, validation and ranking step, is to refine the results from the first step based on the service requirements, the SSDM, and the non-monotonic rules.



CG Similarity Calculation

• A CG similarity *Sim* is calculated through concept nodes similarity S_c and relation node similarity S_r .

$$\begin{split} S_{c} &= 2 \Biggl(\sum_{c \in \bigcup o} (weight(c) \times \beta(\pi_{G_{1}}c, \pi_{G_{2}}c)) \Biggr) \middle/ \Biggl(\sum_{c \in G_{1}} weight(c) + \sum_{c \in G_{2}} weight(c) \Biggr) \\ S_{r} &= \frac{2m(G_{c})}{m_{G_{c}}(G_{1}) + m_{G_{c}}(G_{2})}, \quad Sim = S_{c} \times (a + (1 - a) \times S_{r}) \end{split}$$

- $\bigcup O$ is the union of all of the common generalisation graphs of G_1 and G_2 .
- $\beta(\pi_{G_1}c, \pi_{G_2}c)$ is a function to calculate the semantic similarity between two concepts.
- $m(G_c)$ is the number of the relation nodes in the common overlaps of G_1 and G_2 .
- is the number of the relation nodes of the common overlaps in G_i and the $v_i \in G_i$ aps' adjacent relation nodes.
- *a* is a value between 0 and 1 representing the impact factor of S_r , which make sure that the overall similarity *Sim* will not be 0 unless both S_c and S_r are 0.



Semantic Similarity Ranking

• In the second step of the service discovery, according to the service requirement and the SSDM, the similarities between the services and the requirement are calculated.

 $sim(R, S) = \frac{\sum_{\substack{\forall \alpha \in \lambda}} \omega \times dist(\alpha(R), \alpha(S))}{max(\lambda(R), \lambda(S))}$

- λ : a set of all the semantic characteristics functions.
- λ (): a function that returns the number of semantic characteristics.
- α (): an element of λ that returns a semantic characteristic which can be, e.g. an element of the metadata in the SSDM or the inputs and outputs of a service.
- dist(): a function that calculate the semantic distance between two semantic characteristic and its returned value is between 0 and 1.
- ω : a weight factor that specifies how important a semantic characteristic to a learner is and its value is between 0 and 1.
- *max():* a function returns the greater of its two arguments values.
- R and S: the service requirement and a candidate service.

Semantic Distance Calculation

• Tree Based Similarity

 the semantic similarity between two topics in a ontology is defined as a function of the meaning shared by the topics and the meaning of each of the individual topics.

 $sim(p,q) = \begin{cases} 1 & \text{if } type(p) = type(q) \text{ and } instance(p) = instance(q) \\ depth/(depth+1) & \text{if } type(p) = type(q) \text{ and } instance(p) \neq instance(q) \\ 2d_c/(d_p + d_q) & \text{if } type(p) \neq type(q) \end{cases}$

• Semantic Cosine Similarity

- Two items i_p and i_q are considered as two column vectors in the user requirement matrix. The similarity between items is

measured by computing the cosine of these two vectors.

$$sim(i_p, i_q) = cos(i_p, i_q) = \frac{i_p \bullet i_q}{\sqrt{\|i_p\| * \|i_q\|}}$$

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Conclusion

- A Context-based Semantic Description Framework (CbSDF) is proposed for service description and a two-step service discovery mechanism for service search.
- Aiming to provide a service description framework and a search mechanism that can tolerant imprecise specified service requirements.
- The key technologies used to capture the semantics from imprecise requirements and validate the service discovery results are the CG and the non-monotonic logic, i.e. Defeasible Logic.
- Continue future research on CG and non-monotonic rules in order to improve service description, discovery, and composition techniques.
- Design a suitable evaluation model to evaluate our work.