

Towards a Knowledge Graph for Teaching Knowledge Graphs

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Abstract

This poster paper describes the ongoing research project for the creation of a use-case-driven Knowledge Graph resource tailored to the needs of teaching education in Knowledge Graphs (KGs). We gather resources related to KG courses from lectures offered by the Semantic Web community, with the help of the COST Action Distributed Knowledge Graphs and the interest group on KGs at The Alan Turing Institute. Our goal is to create a resource-focused KG with multiple interconnected semantic layers that interlink topics, courses, and materials with each lecturer. Our approach formulates a domain KG in teaching and relates it with multiple Personal KGs created for the lecturers. Poster Page (GitHub): <https://naiayti.github.io/TeachingKnowledgeGraphs.io/>

Keywords

Educational Knowledge Graph, Domain Model, Teacher Model, Personal Knowledge Graph

1. You ask, we answer: an unorthodox Introduction

Is there really a problem? Teaching courses at the university level can be a demanding task; the field is radically evolving with topic trends changing each year, the material needs to be up-to-date, and finding teaching resources for a topic outside of one's main expertise can be challenging. The challenge arises both from not having a central hub, which can assist in accessing educational resources, topics, and courses, and also from the pluralistic naming and nonuniform format of similar resources across different institutions. Moreover, in the courses offered by the semantic web community, we observe that equivalent courses cover diverse topics, and these courses often highly differ in naming, such as "Web AI" and "Knowledge Engineering". Also, the educational material for lectures and labs is found in a large variety of modalities. From the COST DKG Workshop "Teaching Knowledge Graphs", Malaga, September 2023, it was identified the necessity for providing a resource that could highlight the topics and material taught in KG courses and a platform in which lectures could connect and exchange materials. Therefore, the proposal for a teaching Knowledge Graph (KG) resource was highlighted.

Who cares? Lecturers who teach Knowledge Graph (KG) courses and students who follow these courses can benefit from our resource in multiple use cases, as indicated in Table 1. Moreover, the Semantic Web community needs a methodology for creating an educational

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teaching KG. As quality education is underlined as a Sustainability Goal [1], raises the duty of setting standards for educational Semantic Web applications and KG resources in education. Consequently, we guarantee the quality of our resource by gathering topic descriptions and courses from the experts who teach them.

Ok, but isn't it already done? All we know, we are the first working towards a teaching KG for KG courses. However, our contribution does not rely only on this. As Semantic Web technologies have been widely used in the education domain, ontologies are commonly used for semantically enhancing e-learning applications [2, 3]. Additionally, educational ontologies have been deployed for KG construction [4] and recently, a few educational-purpose KGs [5] have been created for a plethora of applications [6, 7, 8, 9, 10, 11, 12]. Moreover, Personal Knowledge Graphs (PKG) have recently been developed for educational applications [13, 14, 15], where they utilise semantic resources, such as linking entities to encyclopedic KGs. Nonetheless, they are not used in connection to a bigger domain or encyclopedic KG as they do not exchange information, such as updating information about the triples in the encyclopedic or domain KG.

However, none of these approaches goes beyond the factual knowledge representation and limits each resource description to high-level connections, such as which resource is taught first and who is the resource's author.

To the best of our knowledge, we are the first to aim to extract knowledge from each resource, such as topics from educational material, and represent them as new knowledge in the KG. This is significantly important, as one of our main contributions lies in defining a new way of creating educational KGs. We achieve this by building a use-case-driven and resource-focused KG that offers a novel representation of the data it contains. The new representation goes beyond the factual knowledge and extraction of metadata, and further includes the statistical analysis and document information retrieval outcomes as part of the KG entities and properties. An example of this case is applying topic modelling to lecture notes [16], and enhancing the teaching KG with new entities as subtopics extracted from the lecture notes.

Fine, but how are you going to make it? We reuse Semantic Web vocabularies, and standards to develop the teaching KG [17]. Our resource aims to interlink topics, courses, and educational material in order to enable teachers to enhance their courses and help students

Table 1: The use cases presented as competency questions

Competency Questions
For (sub)topic X
Who is teaching X?
Which are the materials for X?
Which are the prerequisites of X?
Which are the laboratories for X?
For course Y
Who is the target audience for Y?
Which educational level does Y target?
Who is teaching Y?
Which slides are linked to Y?
Which labs are part of Y?
Which courses are similar to Y?
Which are suggested resources for Y?
For dataset D
Which exercises exist for D?
Which courses use D?
For material M
Which courses use a material similar to M?
How much similar is M to another material?
Which topics does M cover?
Is M open access?

access multiple resources relevant to their learning goals. Driven by the needs of the users and adding granularity to the representation of material in our teaching KG resource, we aim to contribute both to the Semantic Web advancements as well as set the basis for advanced learning analytics applications.

The teaching KG consists of two parts, the domain model and the user model, which are developed as extensions of the EduCOR ontology [2]. At first, we create a domain model, the teaching KG resource, which contains the topics taught, the educational material, such as the lecture notes and laboratories, and the courses as sequences of the material. Moreover, we model each lecturer as a user model via PKGs [14] which have been shown capable of enabling smart learning environment analytics [18].

Aren't there any limitations? There are plenty of limitations to consider when building teaching KGs. Firstly, not all materials are open source and accessible, as frequently educational content stays hidden behind paywalls or requires institutional access. To address this, we develop the PKGs so the users can access the specific materials from each instructor and gain access to that knowledge. Moreover, the diversity in educational material and non-standardized course descriptions make automation the biggest challenge. The material and courses are present in various formats, which make their integration, interlinking, and maintenance a non-trivial task that we aim to tackle by parsing the text that can be extracted from each resource using symbolic and sub-symbolic techniques [19].

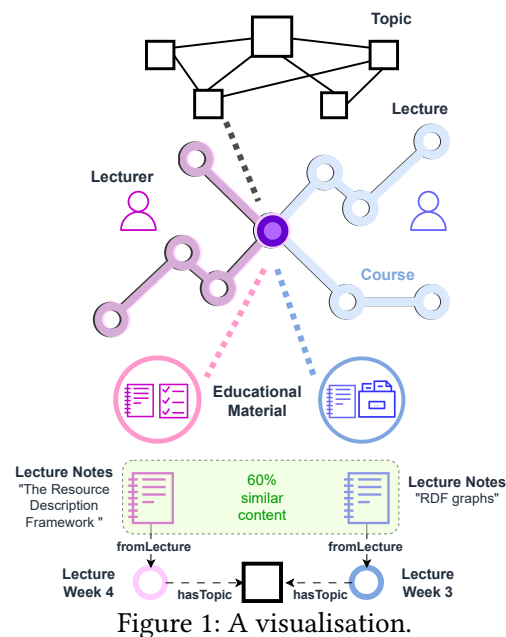


Figure 1: A visualisation.

2. Can you give me an example?

In Figure 1, we provide a visualisation of the multiple layers (topic, lecture, lecturer, course, educational material) present in the teaching KG, in the top part of the Figure, and a paradigm for assisting content retrieval based on semantic similarity, in the bottom part. For the latter, we follow techniques similar to literature [10, 20] to compute the semantic similarity between resources which enables to classify and access similar resources via the connections to the topics ontology.

Moreover, we foresee a hierarchical structure in the topics level. As topics can be classified to general as hypernyms, such as “Data Representation”, and more specified as subtopics, such as “RDFS representation”. We can introduce the relation of prerequisite among topics via statistical analysis of the available resources. The prerequisites are extracted by the course order of lectures and the topics those materials cover. For example, the topic of RDF based on statistical analysis of the gathered material it would be a prerequisite of the SPARQL topic.

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