

Knowledge Graph Embedding with Transfer Learning on Biterm Topic Model of Entity Description

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Abstract

Knowledge graph embedding aims to encode both entities and relations into a continuous low-dimensional space. Most methods concentrate on the structure information of the knowledge graph whose nodes represent the entities and edges represent the relations between entities. In fact,there are usually concise text for entities,which can be modeled with topic model and well utilized by existing embedding methods. In this paper,we propose a novel method for knowledge graph embedding by uncovering the topics within the short entities descriptions and transferring the topic model to entity embedding. Specifically,we learn the topics by directly modeling the generation of word co-occurrence patterns in the whole entity description and transfer the topic model of the entity description to the knowledge embedding with transfer learning. We evaluate our method on two tasks,including knowledge graph completion and entity classification. Experimental results on Freebase show that,by special modeling the topic of the entity description and transfer learning,our method can significantly outperform state-of-the-art method.

1 Introduction

Hello World.

2 Related Work

Hello World.

2.1 TransE

Given a training set D of triples (h,r,t) composed of two entities h,t in E (the set of entities) and a relationship r in R (the set of relationships),TranE

learns the vector embeddings of the entities and the relationships,by minimizing a margin-based ranking criterion over the training set:

2.2 Biterm Topic Model

2.3 Tranfer Learning

Hello World.

3 Our Approach

In this section,we introduce the transfer learning based TransE(TransT) that learns representations of entities and relations utilizing the topic model of short entity description. In TransE and TransT,we have entity set E and relation set R ,and learn to encode both entities and relations in R . Given a KG represented by a set of triples $S=(h,r,t)$ with each triple composed of two entities $h,t \in E$ and their relation $r \in R$,there are also concise text for most entities in the KG. We learn the topics by directly modeling the generation of entity co-occurrence patterns in the entities description corpora by Biterm topic model and transfer the topic model of the entity description to the knowledge embedding with transfer learning method.

3.1 Transferring the Biterm Topic Model to Entities Embedding

From the task of Biterm topic model,we learn the topic distribution of entity description which can represent the semantic of the entities well.Each topic model of the entity can transfer to its embedding by the feature representation transferring method.

$$h^* = h + \lambda \cdot o, t^* = t + \lambda \cdot o \quad (1)$$

$$L = \sum |h^* + r - t^*| \quad (2)$$

3.2 Training the Whole Knowledge Graph

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4 Experiments

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5 Conclusion and Future Work

It is also advised to supplement non-English characters and terms with appropriate transliterations and/or translations since not all readers understand all such characters and terms. Inline transliteration or translation can be represented in the order of: original-form transliteration “translation”.

Acknowledgments

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