SSP: Semantic Space Projection Knowledge Graph Embedding with Text Descriptions

Han Xiao, Minlie Huang, Lian Meng, Xiaoyan Zhu

State Key Lab. of Intelligent Technology and Systems,
National Lab. for Information Science and Technology,
Dept. of Computer Science and Technology, Tsinghua University, Beijing 100084,
PR China

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Problem Definition To incorporate the textual descriptions with the fact triples.

Artificial intelligence is the intelligence exhibited by machines or software.

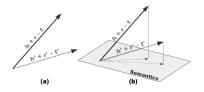
(Artificial Intelligence, Subdisciplines, Machine Learning)

A scientific discipline that explores the construction and study of algorithms that can learn from data...

- Motivations
 - Discovering semantic relevance between entities.
 - Offering precise semantic expression.
- Related Work could not characterize the correlations.
 - ▶ Jointly: $\mathbf{w} = \mathbf{e}$.

Methodology: Projecting the embedding procedure onto a semantic hyperplane.

$$f_r(h, t) = -\lambda ||\mathbf{e} - \mathbf{s}^{\top} \mathbf{e} \mathbf{s}||_2^2 + ||\mathbf{e}||_2^2$$



- Semantic Vector Generation: Topic Model.
- Objectives.

$$\mathcal{L} = \mathcal{L}_{embed} + \mu \mathcal{L}_{topic}$$

$$\mathcal{L}_{embed} = \sum_{\substack{(h, r, t) \in \Delta \\ (h', r', t') \in \Delta'}} [f_{r'}(h', t') - f_{r}(h, t) + \gamma]_{+}$$

$$\mathcal{L}_{topic} = \sum_{e \in E, w \in D_{e}} (C_{e,w} - \mathbf{s}_{e}^{\top} \mathbf{w})^{2}$$
(1)

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- Correlation Perspective: There exists the important restriction, that the entities co-occur in a triple should be embedded in the semantic space composed by the associated textual semantics.
- ➤ **Semantic Perspective**: Our model characterizes the strong correlations with a semantic hyperplane, which is capable of taking the advantages of both two semantic effects.
 - Semantic Relevance.
 - Precise Semantic Expression.

Experiments: Knowledge Graph Completion.

FB15K	Mean Rank		HITS@10	
TransE	210	119	48.5	66.1
TransH	212	87	45.7	64.4
Jointly	167 ¹	39 ¹	51.7 ¹	77.3 ¹
DKRL(BOW)	200	113	44.3	57.6
DKRL(ALL)	181	91	49.6	67.4
SSP (Std.)	154	77	57.1	78.6
SSP (Joint)	163	82	57.2	79.0
WN18	Mean Rank		HITS@10	
TransE	263	251	75.4	89.2
TransH	401	338	73.0	82.3
SSP (Std.)	204	193	81.3	91.4
SSP (Joint)	168	156	81.2	93.2

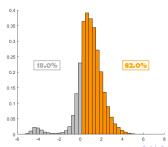
Experiments: Entity Classification.

Metrics	FB15K	FB20K
TransE	87.8	-
BOW	86.3	57.5
DKRL(BOW)	89.3	52.0
DKRL(ALL)	90.1	61.9
NMF	86.1	59.6
SSP (Std.)	93.2	-
SSP (Joint)	94.4	67.4

Semantic Relevance Analysis

	$SSP(S.)_{\# \leq 100}$	$SSP(J.)_{\# \leq 100}$
E _{#≥500}	601	672
E #≥1000	275	298
E _{#≥2000}	80	89
E _{#≥3000}	32	39
E _{#≥5000}	3	3

▶ Precise Semantic Expression Analysis



Conclusion.

- In this paper, we propose the knowledge graph embedding model SSP, which jointly learns from the symbolic triples and textual descriptions.
- SSP could interact the triples and texts by characterizing the strong correlations, by which means, the textual descriptions could make more effects to discover semantic relevance and offer precise semantic expression.
- Extensive experiments show our method achieves the substantial improvements against the state-of-the-art baselines.

Thanks.

Summaries

- 1. From the geometric perspective of **DATA**, we propose *TransG* to model multiple relation semantics.
- From the geometric perspective of MODEL, we propose ManifoldE to achieve an algebraic well-posed system and a flexible geometric form.
- From the geometric perspective of INTERACTION between texts and triples, we propose SSP to utilize textual descriptions.

Resources and Handsome Author





XiaoHan 肖寒 Home Page: http://www.ibookman.net

Thanks for your attention.

