ESA: Entity Summarization With Attention





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Introduction

Query a subject named *Hagar Wilde* in DBpedia with SPARQL:

RDF triples: <Subject, Predicate, Object>

Search Hager Wilde in DBpedia: < Hagar Wilde, Predicate, Object>

Total 65 RDF triples

Virtuoso SPARQL Query Editor

Default Data Set Name (Graph IRI)

http://dbpedia.org

Query Text

Select distinct ?Predicate ?Object where
{
<http://dbpedia.org/resource/Hagar_Wilde>
?Predicate
?Object
}

Predicate	
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://www.w3.org/2002/07/owl#Thing
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://xmlns.com/foaf/0.1/Person Object
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://dbpedia.org/ontology/Person
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://www.ontologydesignpatterns.org/ont/dul/DUL.owl#Agent
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://www.ontologydesignpatterns.org/ont/dul/DUL.owl#NaturalPerson
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://www.wikidata.org/entity/Q215627
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://www.wikidata.org/entity/Q24229398
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://www.wikidata.org/entity/Q28389
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://www.wikidata.org/entity/Q36180
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://www.wikidata.org/entity/Q5
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://dbpedia.org/ontology/Agent
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://dbpedia.org/ontology/ScreenWriter
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://dbpedia.org/ontology/Writer
http://www.w3.org/1999/02/22-rdf-syntax- ns#type	http://schema.org/Person

Introduction

Select Top-5 triples from total 65 triples mentioned above:

```
<a href="http://dbpedia.org/ontology/based0n"> <a href="http://dbpedia.org/resource/Hagar_Wilde"> http://dbpedia.org/ontology/birthDate</a> "1905-07-07"^^ <a href="http://www.w3.org/2001/XMLSchema#date"> http://dbpedia.org/ontology/deathDate</a> "1971-09-25"^^ <a href="http://www.w3.org/2001/XMLSchema#date"> http://www.w3.org/2001/XMLSchema#date</a> <a href="http://www.w3.org/2001/XMLSchema#date</a> <a href="http://www.w3.org/2001/XMLSchema#date</a> <a href="http://www.w3.org/2001/XMLSchema#date</a> <a href="http://www.w3.org/2001/XMLSchema#date</a> <a href="http://www.w3.org/2001/XMLSchema#date</a> <a href="http://www.w3.org/2001/XMLSchema#date</a> <a href="http://www.w3.
```

Table 1. Top-5 predicate-object pairs of subject Hagar Wilde

Predicate	Object	_
Name	Hagar Wilde	_
Туре	Women Televison Writers	\rightarrow
Birthdate	1905-07-07	
Deathdate	1971-09-25	_



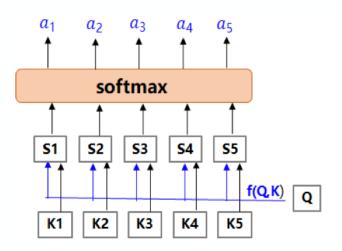
ESA: Machine Attention

Attention Mechanism [1]



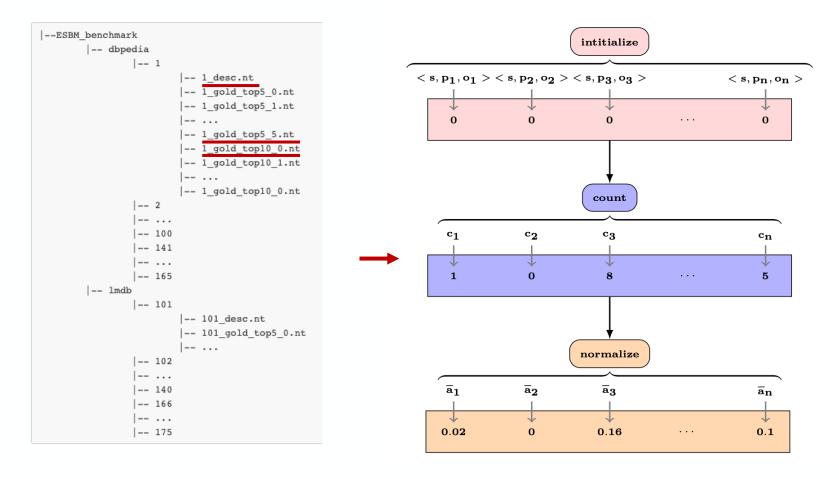
Construct machine attention vector [2]

$$f(Q,K) = \begin{cases} Q^{T} K_{i} & dot \\ Q^{T} W_{a} K_{i} & general \\ W_{a}[Q,K_{i}] & concat \\ V^{T}_{a} tanh(W_{a}Q+U_{a}K_{i}) & perceptron \end{cases}$$



ESA: Gold Attention

Construct gold attention vector



ESA: Model Architecture

Architecture

Attention Mechanism

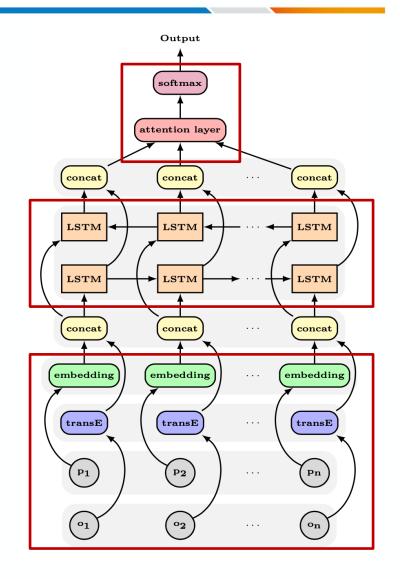
$$\alpha = softmax(\boldsymbol{h}_{S}^{T}\boldsymbol{h})$$
$$\boldsymbol{h}_{S_{i}} = [\boldsymbol{h}_{L_{i}}, \, \boldsymbol{h}_{R_{i}}]$$

Bidirectional Network

$$\mathbf{h}_{L_i} = LSTM_L(x_i, \mathbf{h}_{L_{i-1}})$$

$$\mathbf{h}_{R_i} = LSTM_R(x_i, \mathbf{h}_{R_{i-1}})$$

- Knowledge Representation
 - TransE (Predicates)
 - Word Embedding (Objects)



Experiment

ESBM Datasets: DBpedia & LMDB

	DBpedia		LinkedMDB		ALL	
	k=5	k=10	k=5	k=10	k=5	k=10
RELIN [3]	0.242	0.455	0.203	0.258	0.231	0.399
DIVERSUM [4]	0.249	0.507	0.207	0.358	0.237	0.464
CD [5]	0.287	0.517	0.211	0.328	0.252	0.455
FACES-E [6]	0.280	0.485	0.313	0.393	0.289	0.461
FACES [7]	0.270	0.428	0.169	0.263	0.241	0.381
LinkSUM [8]	0.274	0.479	0.140	0.279	0.236	0.421
ESA	0.310	$\boldsymbol{0.525}$	0.320	0.403	0.312	0.491

Table 1. Experimental Results on ESBM benchmark v1.1 of F-measure

	DBpedia		LinkedMDB		\mathbf{ALL}	
	k=5	k=10	k=5	k=10	k=5	k=10
RELIN [3]	0.342	0.519	0.241	0.355	0.313	0.466
DIVERSUM [4]	0.310	0.499	0.266	0.390	0.298	0.468
CD [5]	-	-	-	-	-	-
FACES-E [6]	0.388	0.564	0.341	0.435	0.375	0.527
FACES [7]	0.255	0.382	0.155	0.273	0.227	0.351
LinkSUM [8]	0.242	0.271	0.141	0.279	0.213	0.345
ESA	0.392	$\boldsymbol{0.582}$	0.367	$\boldsymbol{0.465}$	0.386	0.549

Table 2. Experimental Results on ESBM benchmark v1.1 of MAP

Conclusion

- Neural network is applied into entity summarization task.
- A novel pattern is designed to construct machine attention vectors for modelling supervised attention mechanism.
- Both F-measure and MAP achieves a competitive level in ESBM benchmark v1.1.

References

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Thanks! Q&A.