

# The Open Knowledge System for TAC KBP 2017

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## Abstract

This paper describes the Open Knowledge System for TAC KBP 2017. It is a system that can automatically generate a knowledge base (KB) from a text. The system is based on a deep learning architecture, which consists of a neural network (NN) and a knowledge graph (KG). The NN is used to extract the entities and relations from the text, and the KG is used to store the extracted information. The system is evaluated on the TAC KBP 2017 dataset, and the results show that it can achieve a high performance.

## 1 Introduction

This paper describes the Open Knowledge System for TAC KBP 2017. It is a system that can automatically generate a knowledge base (KB) from a text. The system is based on a deep learning architecture, which consists of a neural network (NN) and a knowledge graph (KG). The NN is used to extract the entities and relations from the text, and the KG is used to store the extracted information. The system is evaluated on the TAC KBP 2017 dataset, and the results show that it can achieve a high performance.

- Entity: a
- Surface Form: a (" ") a
- Entity: ( ) a
- Entity: A: a
- Surface Form: S: a
- Surface Form: B: S (B) a
- Surface Form: F: a
- Surface Form: T: a
- Surface Form: D: a
- Surface Form: S: 3, P: a
- Surface Form: S: 4
- Surface Form: S: 5
- Surface Form: T: 6,
- Surface Form: S: 7 a
- Surface Form: S: 8 a
- Surface Form: S: 9. F a ,
- Surface Form: S: 10 a

## 2 The System Architecture

The system architecture is shown in Figure 1. It consists of a neural network (NN) and a knowledge graph (KG). The NN is used to extract the entities and relations from the text, and the KG is used to store the extracted information.

a      a      a

a a S a C NLP (Ma  
a ., 2014); ) a a  
a - a ; )  
a a a ; ) a a  
a a :  
1) a a a a  
a , . .  
< > a , a  
a a .  
2) a a a a  
a a a .

#### 4 The Event Nugget Detection and Co-reference Module

T a  
a , . . , E D  
a C a a , R a C a a a E  
C R . A a  
a LDC2015 DEFT R ERE E  
T a A a \_R2\_V2 a a .

##### 4.1 Event Detection and Classification

O a a a . T  
a a a .  
1)  
2) a  
3) a  
4)  
W a CNN- a a a  
a a a . W P-O-S  
a . A  
a , a  
a a .

##### 4.2 Realis Classification

W CNN- a a a  
4.1 R a a  
(ACTUAL,GENERIC,OTHER). E  
a a , a a  
a  
.

#### 4.3 Event Coreference Resolution

W a  
a . W  
a a  
.

#### 5 The Relation Extraction Module

T a a  
, a , RNN- a a  
a , O IE- a a  
a I R a E a .

##### 5.1 RNN-based Relation Extraction

O a a a a  
B a GRU a  
a a . T a  
a a , a a  
a . A a  
a a -GRU  
. W a a a  
T a a -GRU  
a a - a a  
a a a a  
a a a .

##### 5.2 OpenIE-based Relation Extraction

T a O IE a  
a ( a a O IE  
V4), a a  
a a .  
F , a  
a a  
O I a E a (O IE)  
U Wa  
(UW). A O IE a  
a a a ,  
a 2) . B a O IE  
O . W a O  
a a a  
a . I a a a  
a a , a  
a a . W a  
a a  
a TAC a . F a  
a , Ba a O a a a : T U.S.

T a a . a T , a a  
a : (Ba a O a a,  
U.S.) a (Ba a O a a, a , )  
a (Ba a O a a, a , T a )  
a (Ba a O a a, a , a ( . . 1, , 2) a ,  
). T TAC, a a  
a a . T a  
a a (A, B, C) A a B a a , a  
C a a , a F a a a a ( . . ,  
O IE a a , - - ),  
a a a . N , a . F a a a  
A a C a a E a a ( . , : a ),  
D . I a - , 10 a a  
a a a a . a  
B , a B 41 a  
a a a a ,  
a a a a a  
T a , a 1230864 a  
. A a , a  
16792 . D ,  
a O IE a .

### 5.3 Implicit Relation Extraction

A a a a a  
a I R a I a E a , F a a a a a  
a a a a . T a a  
(S a a , 2015), , IMPLIE  
a a , a  
a C a O IE, a  
a . N , a a ,  
a " T " 41 a . F  
" : ", a a a  
E a , a a a 41  
a , a a  
a a O IE. A  
a a a a a  
a a O IE  
IMPLIE . T , IMPLIE a  
.

### 5.4 The Combination Strategy

T a a  
a . I a  
a a a ,  
a a a -  
.

## 6 The Sentiment Module

W a SGD a a  
. F , a  
IMDB  
a a T a a  
a . T  
a a SGD a a  
. S a , a a a  
a a a a a  
a a . F N a a ,  
a a a a a  
A L R a a  
SVC a .

## 7 The Entity Linking and Clustering Module

T a a ,  
a a . F  
W a. T  
a  
.

### 7.1 Query Expansion

I , a a  
a . T  
a a a :  
1) S a . A a  
a  
2) S a W a a a a  
a a . A  
.

3) S a W a a a a  
a a . A

4) Ca a a  
a . A  
a  
a .

## 7.2 Candidate Entity Generation

I  
a a a a  
a a a  
a a a . Na , a S  
7.1 a a a a  
KB ( . . , W a).

## 7.3 Entity Linking

I a a  
. F ,  
a . S , a  
a a a a  
. A a a  
. T a a a  
:

1) E a . T a  
a a a a  
. U a a a a a

2) Na a . Na , a  
a a a  
KB.

3) C a . W K  
a a  
, a a a

4) W a a a  
E a a a  
W a a a  
5) A a , a a  
a a a a a

KB TFIDF  
a a a  
a a a a a

a a a a a  
a a .

6) Ca a a . A a  
a  
7) T . T a

F a a a a a  
W a a a a , a  
a a a  
a , a  
a a . T a  
EDL2016 a a a.

## 7.4 NIL Entity Clustering

F NIL , a  
NIL a a  
F , a a  
a a a a  
a , a a a  
a , a a a  
a a a  
a a a  
A a  
. T a a a  
. T a a  
. A - a  
a a a a  
F a a a a  
a a a . F a ,  
a a a a a a  
I a a a ,  
a a a a

## 8 The Inference Module

T a a S 4,  
a a a

1) R a - a . F a ,  
a a a a  
a " a " a  
" " Ga . S a , " "  
a " a .

2) R a - a . F a ,  
: a \_ \_ , : a \_ \_ a , :a ,  
a

( && a -> a , G .  
 a a a a  
 a ). F a , A  
 2010 a a 78, a A a  
 1932.  
 3) R a a Ta , 2.

Ta 2 R	a	a
A --- B	B --- C	A --- C
		- a - a
	a	- a - a
a a a	a	- a - a a

4) R - a . F  
 a  
 a / / a , a  
 T a , a a a a  
 a a a a .  
 5) R - a . F a , CEO,  
 a , - , a  
 a : - - - ,  
 a : - - .  
 6) R . F a  
 a , a a  
 a S D

## 9 The Post-processing Module

T a  
 F , a -  
 . S a , a  
 , a a .  
 1) T a a  
 a a . F a ,  
 a ( . . ,  
 ),  
 (PER). T a a a  
 S a NER (F a ., 2005).  
 2) S a a a . C a a  
 a a a a  
 XXXX-XX-XX .  
 3) D a a a . F a ,  
 a a a 0 a a a  
 130 .

## 10 Result Evaluation

T a , a a KB a  
 KB a a a a  
 KB a a . T KB  
 a a a Ta  
 2, 3, 4 a 5.  
 F C S a KB Ta ,  
 ,

Ta 3 E				(M A a , E )					
	a						a a		
	P	R	F1	P	R	F1	P	R	F1
1	50.56	13.87	21.76	32.83	9.01	14.13	35.51	9.74	15.29
2	50.56	13.87	21.76	32.83	9.01	14.13	35.51	9.74	15.29
3	64.32	11.31	19.23	42.05	7.39	12.57	45.36	7.97	13.56

Ta 4 S			a (LDC-MAX, E , K3)						
	0 P	0 R	0 F	1 P	1 R	1 F	A P	A R	A F
1	0.0545	0.0380	0.0448	0.0000	0.0000	0.0000	0.0040	0.0275	0.0070
3	0.0755	0.0506	0.0606	0.0000	0.0000	0.0000	0.0606	0.0367	0.0114

## Conclusion

I a , OKS  
C S a KB T a KBP  
2017. T a  
a , a , E  
a , E N D a C  
a a a . T a a a  
a a .

## Acknowledgements

T Na a K R a  
a D P a C a a  
2016YFB1000902, Na a G a F a a  
R a 973 P a C a a  
2014CB340406, a Na a Na a S  
F a C a a 61772501,  
61572473, 61572469, 91646120, a 61402022.

## References

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P TAC-KBP 2016.  
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I a - a a a  
a 43 A a M A a  
C a a L , 363-370.  
Ma a , S M., Ba R., S a S., a  
E O. 2012. O a a a  
a a .I P EMNLP.  
S a S., G J., R Ba , O E - ,  
a Da S. W . 2013. O a  
a KBP a 3 . I  
P TAC-KBP 2013.  
W J, B A, Ya O, a . C  
La a a K Ba E  
M R a E a [J]. 2013.

L H, Z a Z, J a Y, .a. O KN a TAC KBP 2014.  
I P TAC-KBP 2014.  
C X, J a Y, Wa Y, .a., O KN a TAC KBP  
2015. I P TAC-KBP 2015.  
T a a K., K D., Ma C., a S Y.  
2003. F a - Pa - -S Ta a  
. P HLT-  
NAACL 2003, 252-259.  
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KBP C S a S F .I P TAC-  
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D a a .  
W F. a W D. 2007. A a  
W a. P a ACM  
a a , 41-50.  
Ma , C D., M a S a , J Ba ,  
J F , S J. B a , a Da  
M C . 2014. T S a C NLP Na a  
La a P T I P  
52 A a M A a  
C a a L : S D a ,  
. 55-6