







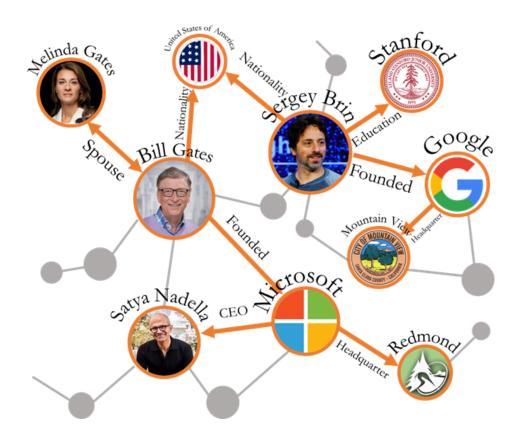
# Learning Relation Entailment with Structured and Textual Information

**Zhengbao Jiang**<sup>1</sup>, Jun Araki<sup>2</sup>, Donghan Yu<sup>1</sup>, Ruohong Zhang<sup>1</sup>, Wei Xu<sup>3</sup>, Yiming Yang<sup>1</sup>, Graham Neubig<sup>1</sup>

Carnegie Mellon University<sup>1</sup>, Bosch Research North America<sup>2</sup>, Ohio State University<sup>3</sup> zhengbaj@cs.cmu.edu

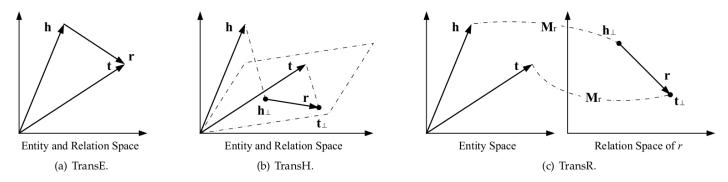
### Motivation

• Relations among entities play a fundamental role in knowledge graphs.



### Motivation

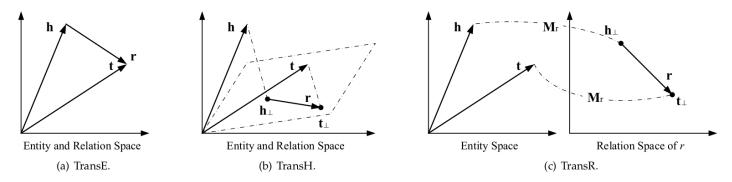
• However, relations are treated as independent.



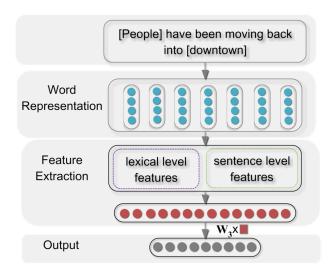
KG embedding: each relation is treated as an atomic unit with separate parameters.

### Motivation

• However, relations are treated as independent.



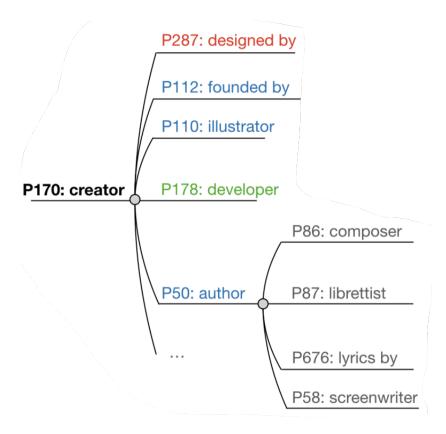
KG embedding: each relation is treated as an atomic unit with separate parameters.



Relation extraction: each relation is an independent class.

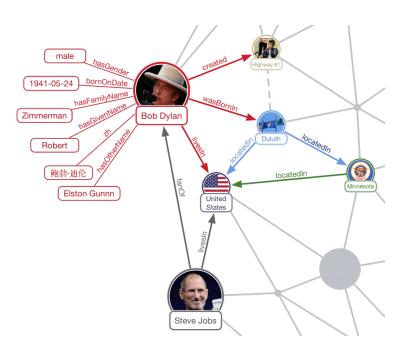
### Meta-relation: Relations Between Relations

• Relation entailment: existence of one relation can entail the existence of another relation.



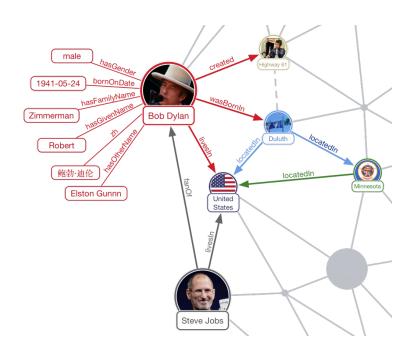
# Applications of Relation Entailment

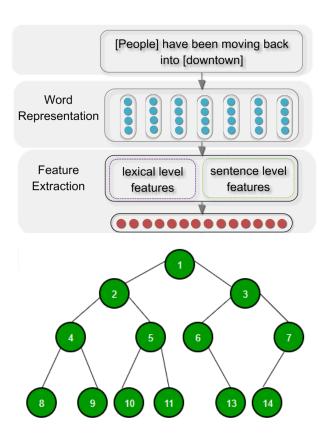
• Knowledge graph representation learning.



### Applications of Relation Entailment

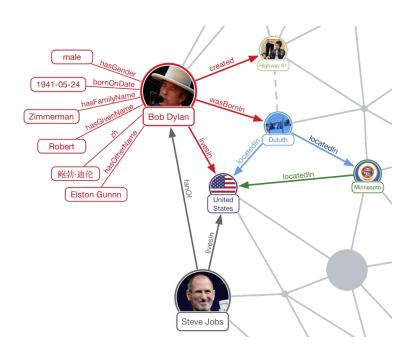
- Knowledge graph representation learning.
- Relation extraction.

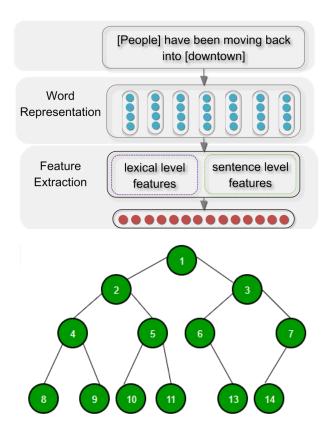


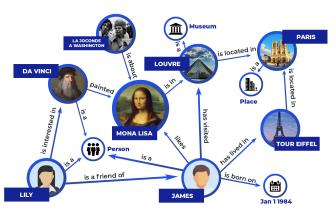


### Applications of Relation Entailment

- Knowledge graph representation learning.
- Relation extraction.
- KG-based question answering.









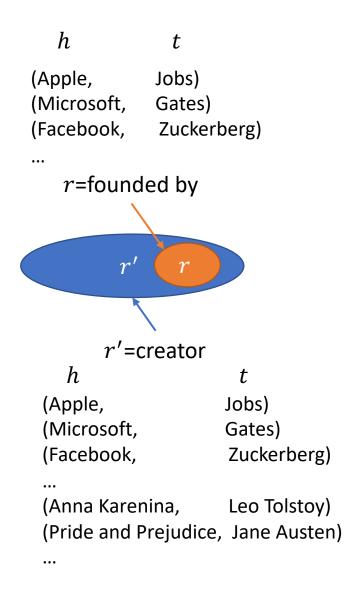
### Relation Entailment Task Definition

```
h t
(Apple, Jobs)
(Microsoft, Gates)
(Facebook, Zuckerberg)
...
r=founded by
```

### Relation Entailment Task Definition

#### Notations

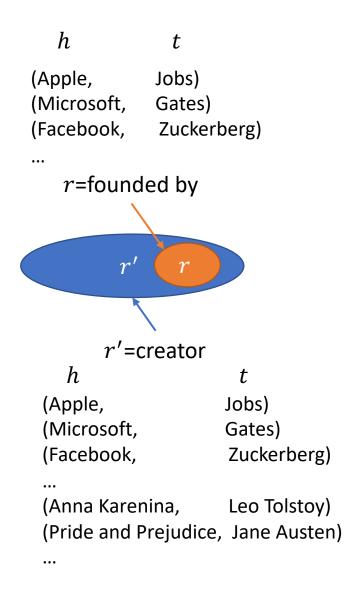
- Head and tail entities  $h, t \in \mathcal{E}$ .
- Relations  $r \in \mathcal{R}$ .
- Instances of a relation  $C_r = \{(h, r, t)^{(i)}\}_i$ .
- Relation entailment
  - $r \models r'$  if and only if  $C_r \subseteq C_{r'}$ .
- Task of predicting relation entailment
  - Given a relation r, choose its (direct) parent  $r' \in \mathcal{L}$ .
  - A  $|\mathcal{L}|$ -way multi-class classification problem.

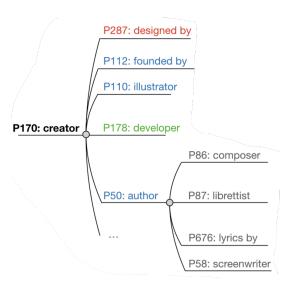


### Relation Entailment Task Definition

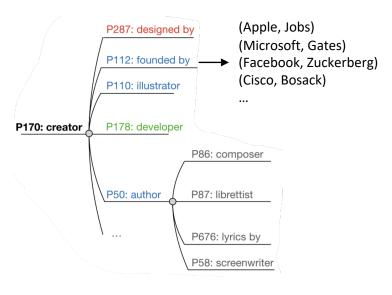
#### Notations

- Head and tail entities  $h, t \in \mathcal{E}$ .
- Relations  $r \in \mathcal{R}$ .
- Instances of a relation  $C_r = \{(h, r, t)^{(i)}\}_i$ .
- Relation entailment
  - $r \models r'$  if and only if  $C_r \subseteq C_{r'}$ .
- Task of predicting relation entailment
  - Given a relation r, choose its (direct) parent  $r' \in \mathcal{L}$ .
  - A  $|\mathcal{L}|$ -way multi-class classification problem.



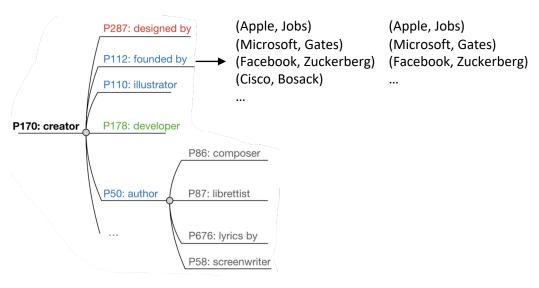


#### 1. Instances collection



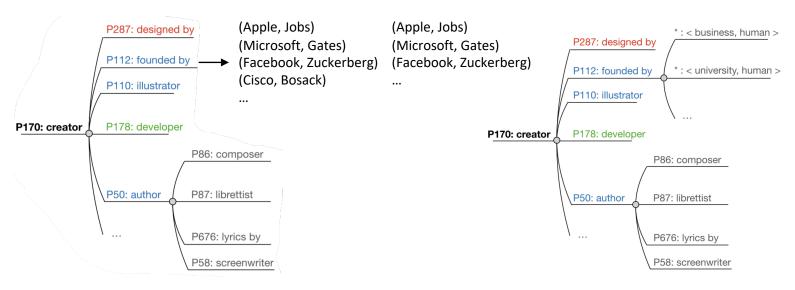
#### 1. Instances collection

### 2. Downsampling



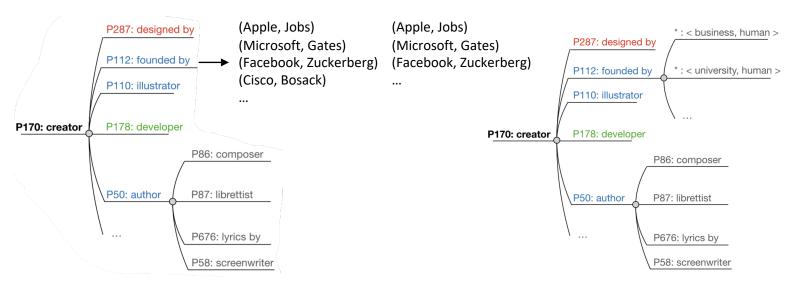
#### 1. Instances collection

### 2. Downsampling 3. Relation expansion



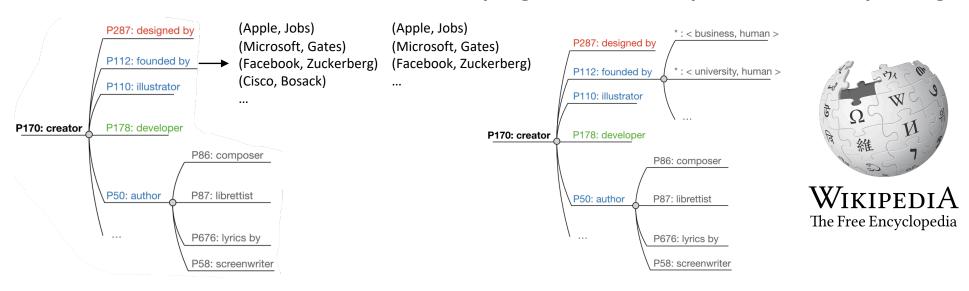
#### 1. Instances collection

### 2. Downsampling 3. Relation expansion



| parent              | Sub-relations   |  |  |
|---------------------|---|--|--|
| parent organization | <pre><laboratory, university="">, <airline, airline="">, <record label="" label,="" record="">,</record></airline,></laboratory,></pre> |  |  |
| architectural style | <railway architectural="" station,="" style="">, <church, architectural="" style="">,</church,></railway>                               |  |  |
| award received      | <film, academy="" awards="">, <human, campaign="" medal="">, <human, scholarship="">,</human,></human,></film,>                         |  |  |

#### 1. Instances collection 2. Downsampling 3. Relation expansion 4. Entity linking

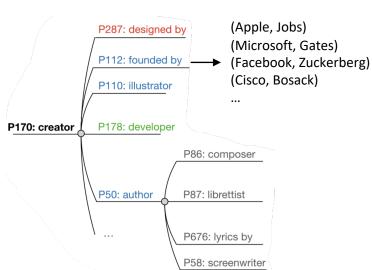


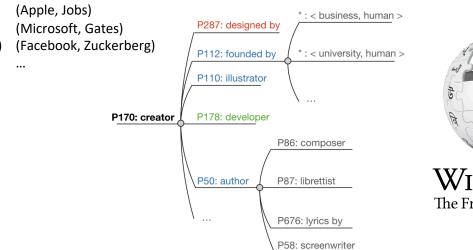
| parent              | Sub-relations   |  |
|---------------------|---|--|
| parent organization | <pre><laboratory, university="">, <airline, airline="">, <record label="" label,="" record="">,</record></airline,></laboratory,></pre> |  |
| architectural style | <railway architectural="" station,="" style="">, <church, architectural="" style="">,</church,></railway>                               |  |
| award received      | <film, academy="" awards="">, <human, campaign="" medal="">, <human, scholarship="">,</human,></human,></film,>                         |  |

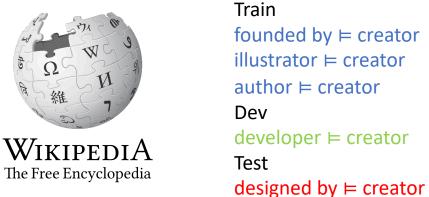
#### 1. Instances collection

### 2. Downsampling 3. Relation expansion 4. Entity linking

#### 5. Train/dev/test split

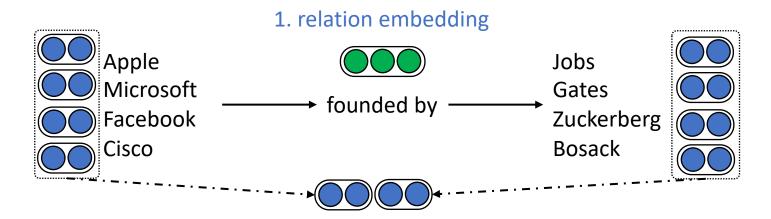




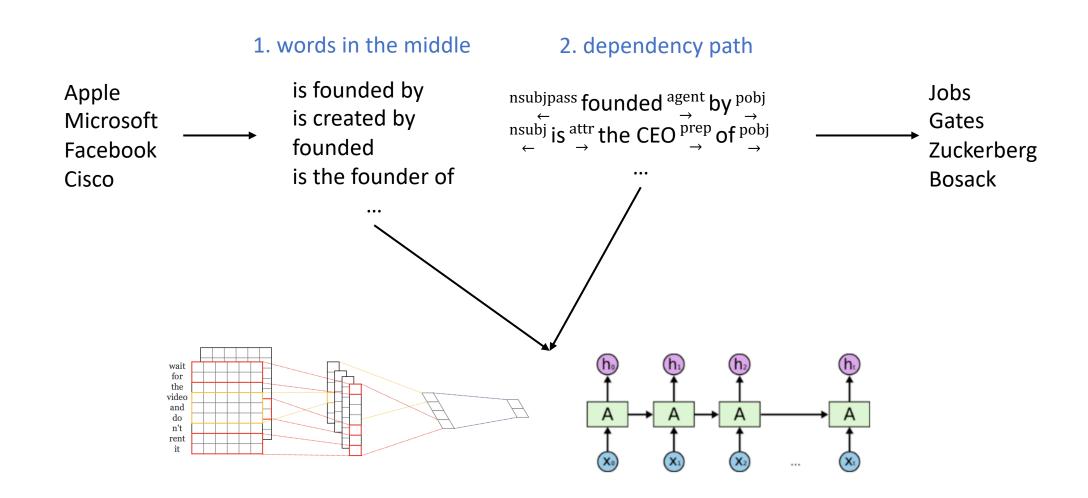


| parent              | Sub-relations  |  |
|---------------------|--|--|
| parent organization | <laboratory, university="">, <airline, airline="">, <record label="" label,="" record="">,</record></airline,></laboratory,> |  |
| architectural style | <railway architectural="" station,="" style="">, <church, architectural="" style="">,</church,></railway>                    |  |
| award received      | <film, academy="" awards="">, <human, campaign="" medal="">, <human, scholarship="">,</human,></human,></film,>              |  |

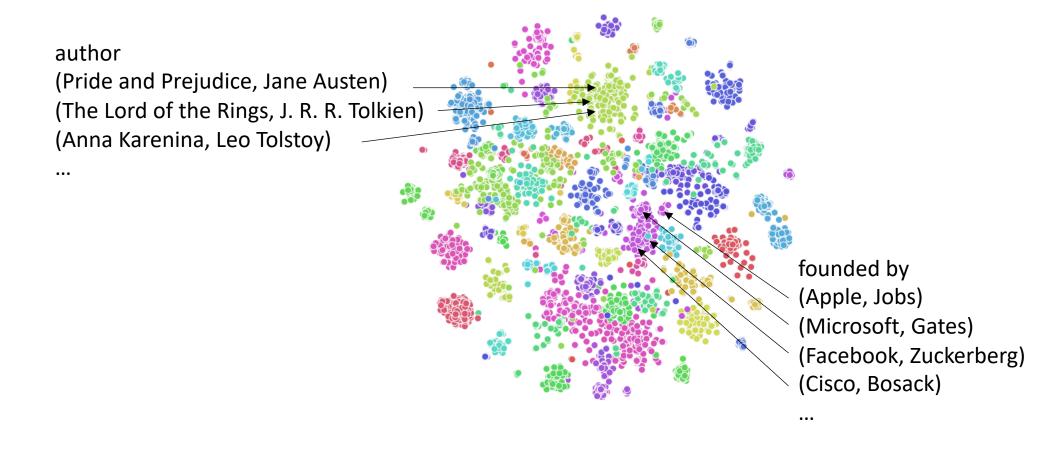
• With structured information



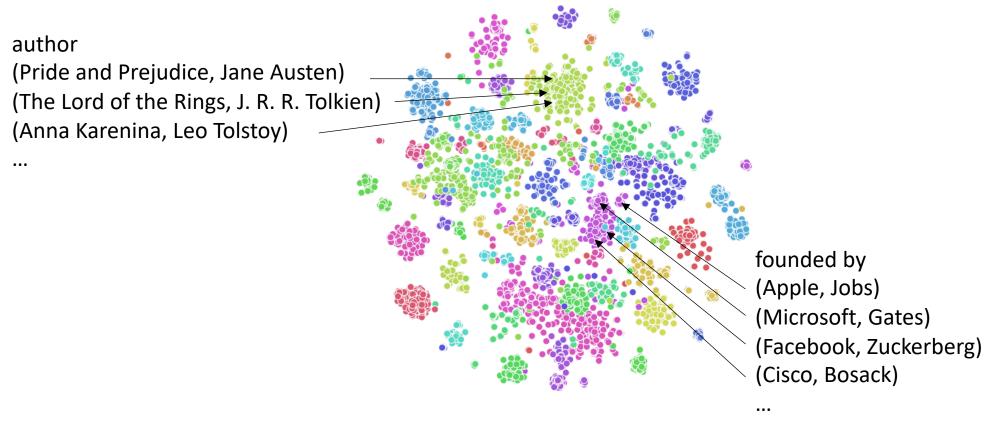
With textual information



• Distribution-based



• Distribution-based



Kernel density estimation with a Gaussian kernel

### Relation Entailment Prediction

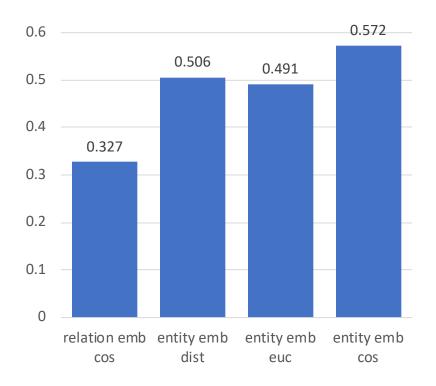
cos( **()**, Unsupervised methods Supervised methods

### **Experimental Settings**

- RelEnt Dataset
  - #Train, #Dev., #Test relations: 2055, 804, 692
  - #Classes: 498

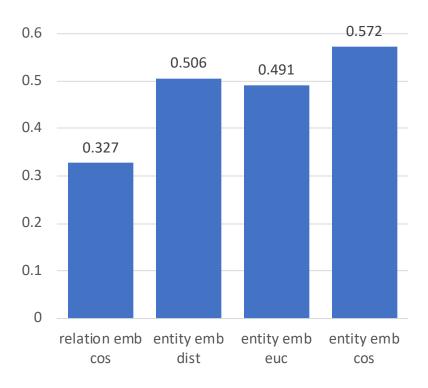
- Evaluation Metrics
  - Accuracy@1, Accuracy@3, and mean reciprocal rank (MRR)

# Unsupervised Methods' Results

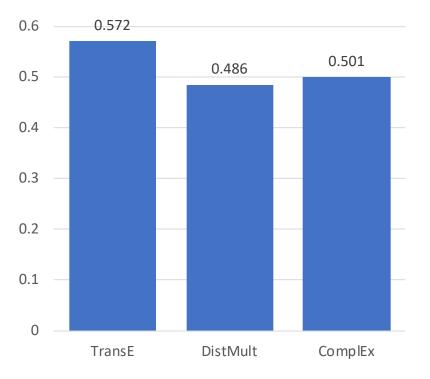


Acc@1 of different unsupervised methods with TransE.

### Unsupervised Methods' Results



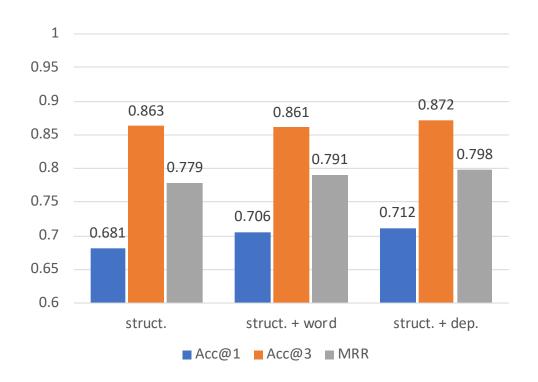
Acc@1 of different unsupervised methods with TransE.



Acc@1 of entity embedding with cosine using different KG representations.

### Supervised Methods' Results

- Supervised > unsupervised.
- Textual information is complementary to structured information.



### Error cases

| Parent      | Child (train) | Child (test)       |
|-------------|---------------|--------------------|
| follows     | has cause     | replaces           |
| instance of | taxon rank    | legal form         |
| participant | performer     | participating team |

(2010 Wimbledon Championships, Roger Federer) (First Continental Congress, George Washington) (Hambach Festival, Ludwig Börne)

> (Runaway, Linkin Park) (The Freewheelin' Bob Dylan, Bob Dylan)

> > (1977 UEFA Cup Final, Juventus FC) (2016–17 Premier League, Watford F.C.) (1956 Wrestling World Cup, Iran)









### Take away

- 1. Both structured and textual information contribute to relation entailment prediction.
- 2. Relation entailment prediction requires high-level abstraction.

Paper: <a href="https://openreview.net/pdf?id=ToTf">https://openreview.net/pdf?id=ToTf</a> MX7Vn

Code: <a href="https://github.com/jzbjyb/RelEnt">https://github.com/jzbjyb/RelEnt</a>