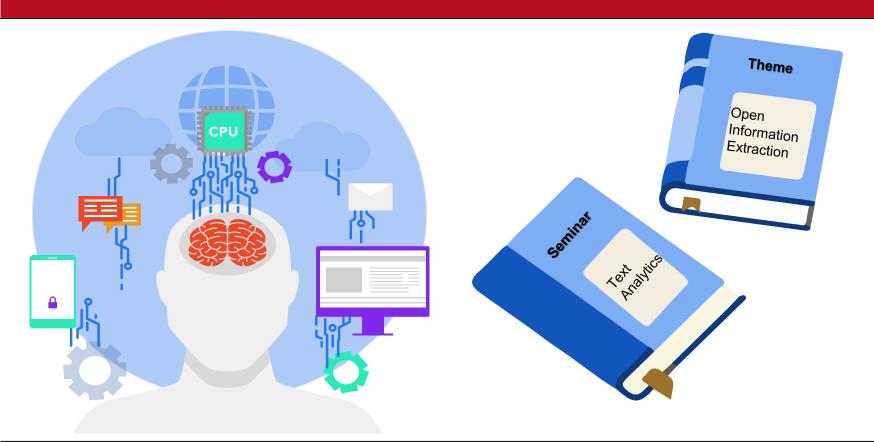
# DocOIE: A Document-level Context-Aware Dataset for OpenIE



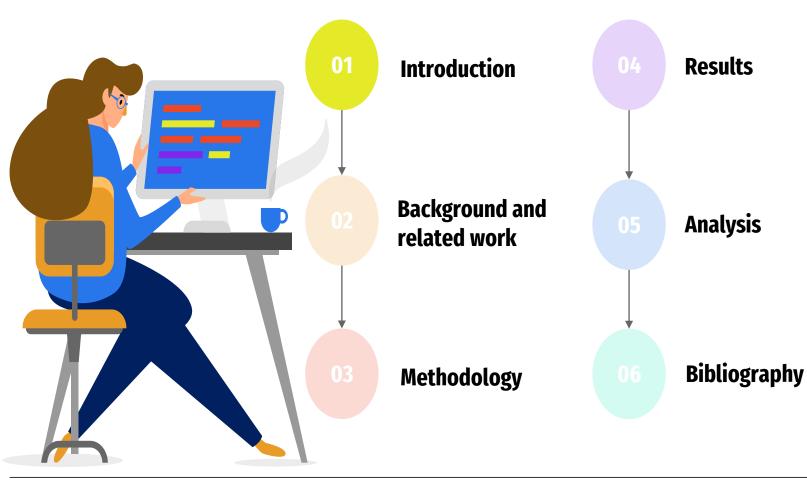
Kuicai Dong, Yilin Zhao, Aixin Sun, Jung-Jae Kim, Xiaoli Li

Presented by : Amira Chebbi







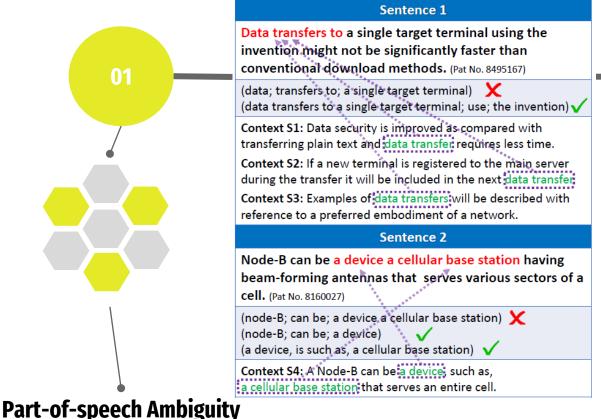


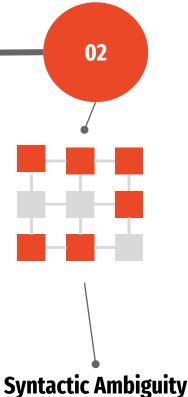
#### **Introduction - Problematic**

Is it a Verb or a Noun?

DocOIE: A Document-level Context-Aware Dataset for OpenIE





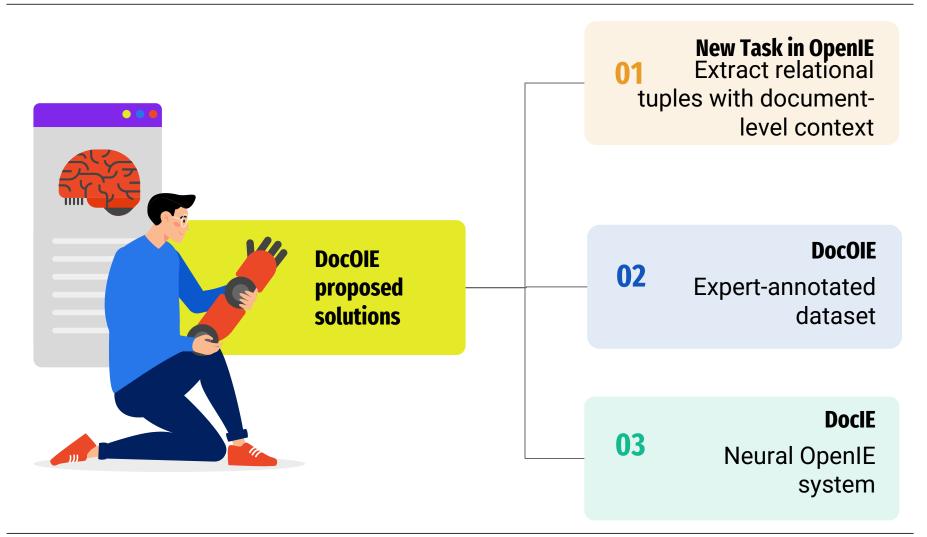


What is the relationship between this two words?

# **Introduction - Proposed Solution**

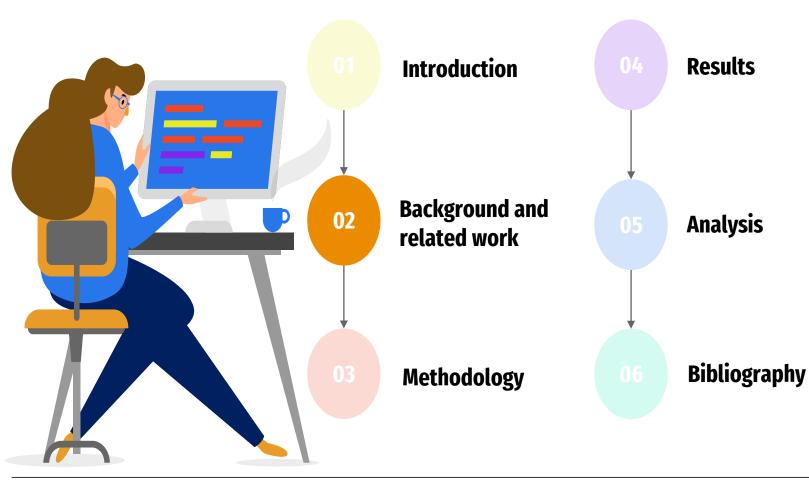
DocOIE: A Document-level Context-Aware Dataset for OpenIE













# **Related work - OpenIE Datasets**



#### **OpenIE Datasets**

OIE2016 (Gabriel Stanovsky and Ido Dagan. 2016)

first large-scale dataset constructed for OpenIE tasks and comes with a standard scoring framework.the gold tuples are automatically generated according to human crafted rules.

Wire57(Lechelle et al., 2019)

manually annotates 57 sentences + scoring improvement

CaRB (Bhardwaj et al. (2019))

50 expert-annotated sentences + a sophisticated scoring framework





# **Related work - OpenIE Datasets - Problems**



#### **Dataset's size problem**

The number of the annotated sentences ramains small

OpenIE Datasets Problems

# No referring to contextual Information

The Annotation is elaborated on a sentence-level and not a document-level





# **Related works - OpenIE Models**



#### **OpenIE Models**

TextRunner (Yates et al., 2007)

neural OpenIE systems (Cui et al., 2018; Zhan and Zhao, 2020; Kolluru et al., 2020a,b)

the first OpenIE system, extract relational tuples based on handcrafted rules or statistical methods

They extract tuples in an endtoend manner, not requiring prior syntactic or semantic analysis.





# **Related works - OpenIE Models - Problems**

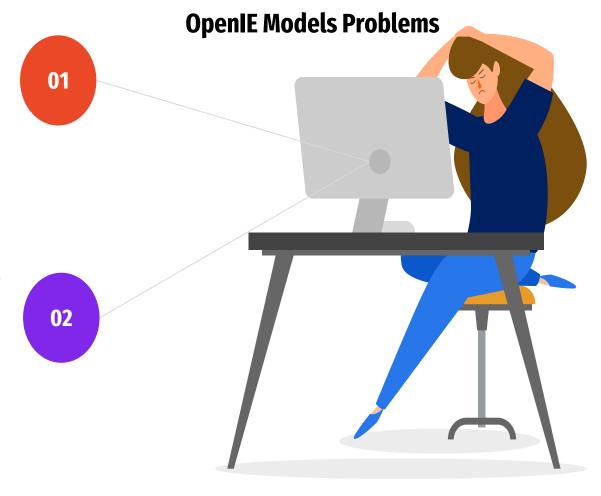


# **Error accumulation for** traditional models

Because they rely on prior syntactic or semantic analysis

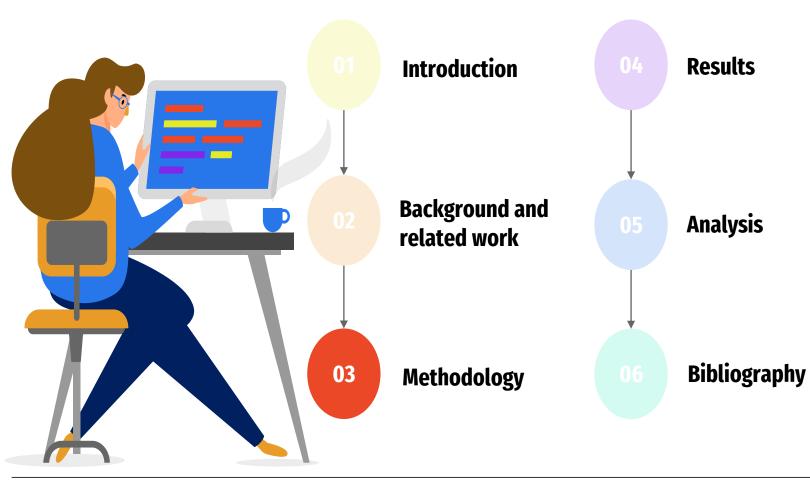
#### No consideration of the documentlevel context

The tuple extraction is based only on the sentence-level









# **Methodology – DocOIE Dataset**



05

#### **Dataset Collection**

Types of documents and document selection

02

#### **Evaluation Dataset**

Selection of the evaluation Dataset

04

#### **Analysis of Evaluation Dataset**

Understand the difficulty of DocOIE

03

#### **Annotation** process

Diffrent stages of the annotation process and the consistency measurement

**Training Dataset** 

Selection of the training dataset

06

#### **Pseudo Label** Generation

Generating pseudo labels for the training Dataset

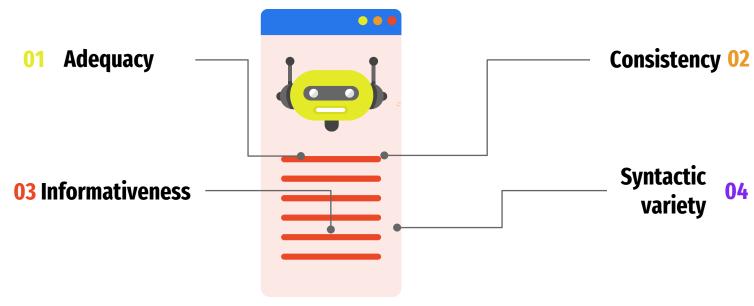


## Methodology –DocOIE Dataset

#### 1- Dataset Collection



- Document types : News and Wikipedia articles
- Document selection criterias :



- Keyword selection criterias : Magnitude and Diversity
- Document cleaning



# Methodology –DocOIE Dataset

#### 2-Evaluation Dataset



- Randomly select 80 documents (40 each domain)
- From each document, select randomly 10 sentences
- Result : 800 expert-annotated sentences from the DocOIE evaluation
   Dataset



# **Methodology – DocOIE Dataset**

# 3- Annotation process



# **Annotation process and consistency measurement**

#### Stage 1

#### 1- Annotation

The two annotators practiced annotations independently on 100 sentences

#### 2- Cross validation

Crossvalidation of the results, discussion and then update

#### Stage 2

#### **1- Annotation**

The two experts annotated independently other 100 sentences 2- Consistency Measurement

Evaluation used at tuple-level using the CaRB scorer

#### Stage 3

Based on the highlevel annotation consistency, each expert annotate independently 300 sentences.



## Methodology –DocOIE Dataset

# 4- Analysis of evaluation dataset



■ Used to understand the difficulty of DoclE, similar to (Gashteovski et al., 2019).

and



#### **Sentence-level Analysis**

Evaluate sentence complexity by the number of :

- Conjunction words
- Terminology mention
- Dependent clause



#### **Tuple-level Analysis**

The analysis of a tuple is based on three points :

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- Negative polarity
- Possibility
- Under-specificity



## Methodology –DocOIE Dataset

# 5- Trainig Dataset





- Select 2400 documents randomly (1200 each Domain)
- The 1200 documents in each domain contain around 120.000 sentences (sufficient for the openIE model)



### Methodology – DocOIE Dataset

# 6- Pseudo Label generation

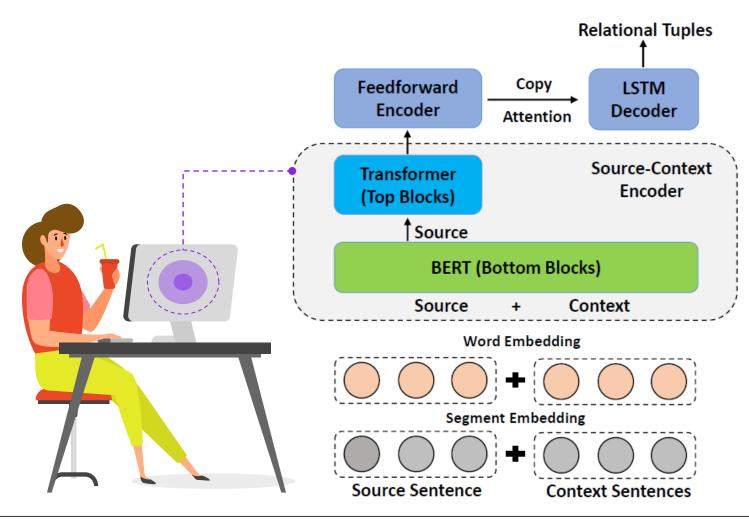


- Generation of the pseudo labels by bootstrapping with traditional OpenIE models(Kolluru et al., 2020b; Cui et al., 2018; Zhao et al., 2020)
- But First: Evaluation of the performance of the traditional OpenIE models on the evaluation Dataset using CaRB scorer( to guarantee better quality of pseudo labels)
- The evaluated models were: Reverb (Fader et al., 2011), Clausie (Corro and Gemulla, 2013), Stanford OpenIE (Angeli et al., 2015), OpenIE4 (Mausam, 2016), OpenIE5, Rev+Oie4 and Oie4+ Rev
- Results shows that : both Reverb and OpenIE4 are the best performing individual models and their combinations lead to the best and second best F1 scores in both domains.



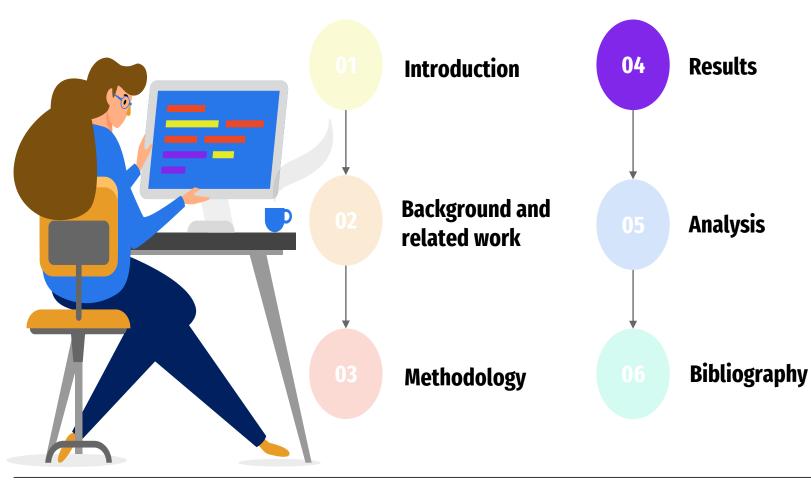
## Methodology – DocIE Model











# Results - DocIE Against sentence-level OpenIE systems



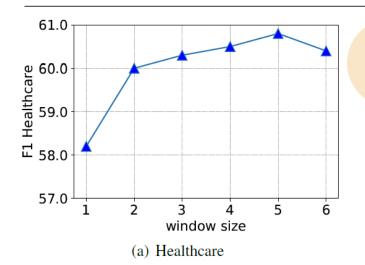
| System                | Healthcare  |      |             |      | Transportation |      |             |             |
|-----------------------|-------------|------|-------------|------|----------------|------|-------------|-------------|
|                       | AUC         | Prec | Rec         | F1   | AUC            | Prec | Rec         | F1          |
| Rev+Oie4              | 36.8        | 75.8 | 47.7        | 58.6 | 31.0           | 74.2 | 42.4        | 54.0        |
| Oie4+Rev              | 35.8        | 59.6 | 55.3        | 57.4 | 30.1           | 53.4 | 52.7        | 53.0        |
| CopyAttention+BERT    | 46.8        | 77.9 | 48.6        | 59.8 | 38.3           | 55.3 | 56.9        | 56.1        |
| IMOJIE                | 39.7        | 80.1 | 46.4        | 58.7 | 35.8           | 63.5 | 49.2        | 55.5        |
| DocIE w/o transformer | <u>47.1</u> | 76.2 | 49.9        | 60.3 | 38.5           | 55.8 | <u>57.0</u> | <u>56.4</u> |
| DocIE w transformer   | 47.4        | 74.4 | <u>51.3</u> | 60.8 | 38.5           | 56.0 | 57.5        | 56.9        |

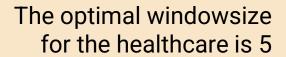
- DocIE with transformer achieves the best AUC and F1 in both domains.
- DocIE without transformer is the second best performer and outperforms all the sentence-level systems.



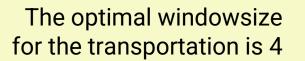
# **Results - Impact of the context Window size**







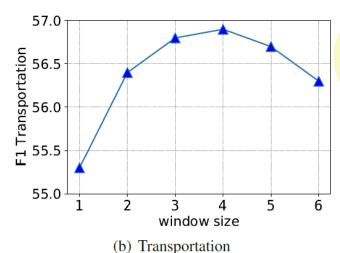




and

Large window size may introduce noise

21



# **Results - Error Analysis**



#### **Incompleteness**

Fails to cover at least one key phrase in either arguments or relations

#### 02 **Incorrect Boundary**

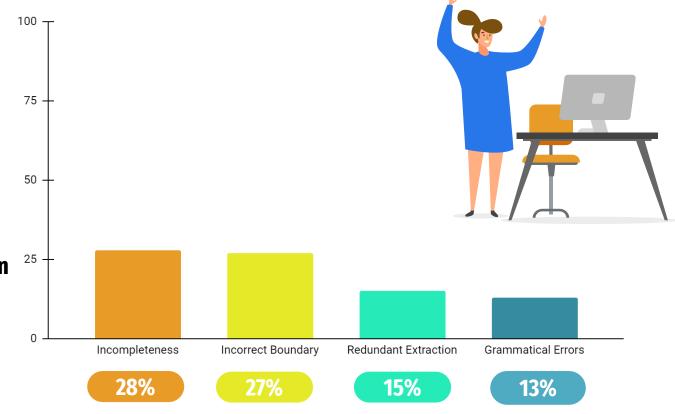
Misinterpretaion of the syntactic meaning of the sentence, leading to incorrect Boundary

#### **Redundant extraction** 03

The same relational fact is extracted multiple times

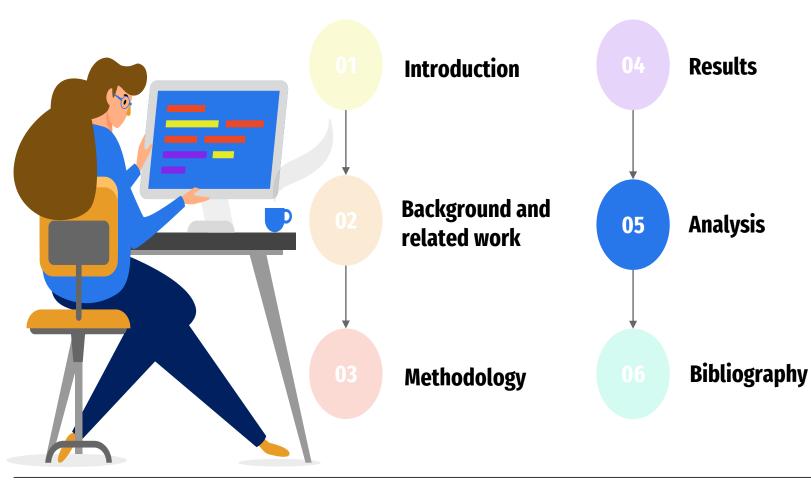
**Grammatical Errors** 04

> Extracions are not grammatically correct









# **Analysis – Contributions**



# **Contribution**facet



**Positive outcomes** 

Better correct understanding of various topics First documentlevel context-aware OpenIE dataset and model

**Relation to public** policy

better analysis and helps in writing the adequate public policy and taking the right measurement in many areas.

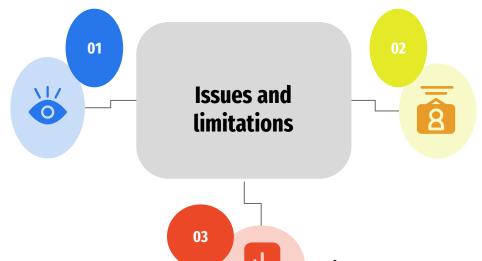


# **Analysis – Possible Issues and limitations**



#### **No Generalization**

DocIE Dataset is restricted to specific type of documents



#### **Domains restriction**

Foucus only on two domains:
Transportion and Health-care

### **Using CaRB scorer for Evaluation**

During the penalization CaRB scorer does not take the contextual sentences into consideration



# **Analysis – Future Work**



#### **Model Improvement**

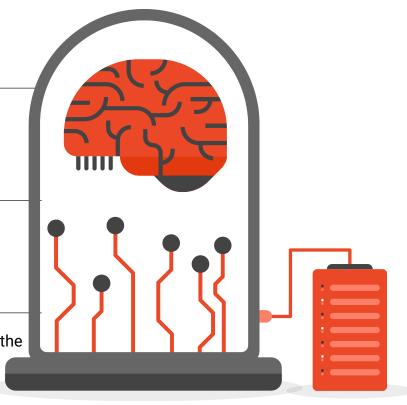
Research on more effective context-aware models

#### **Pseudo-Labels** 02

Invistigate the possiblity of not relying on pseudo labels

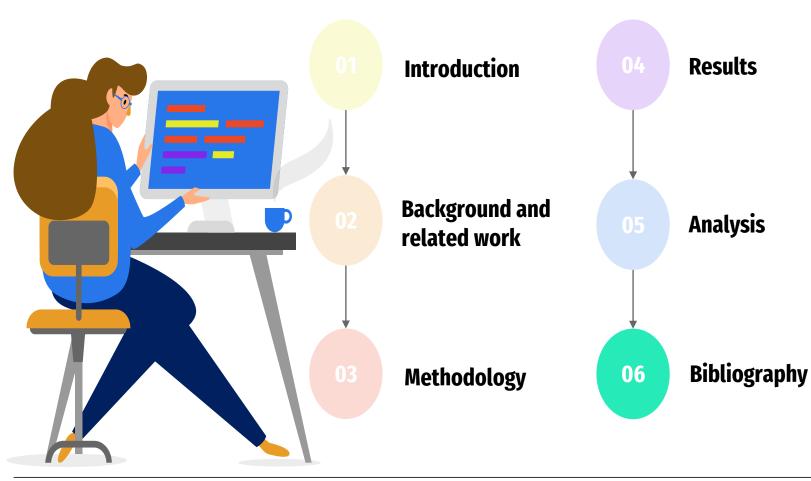
#### **Adapted Scorer** 03

Trying to develop a new scorer that takes into consideration the contextual informations









#### Refrences

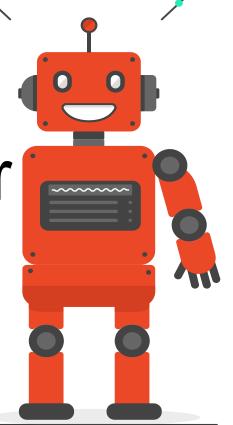


- DocOIE: A Document-level Context-Aware Dataset for OpenIE
- Creating a Large Benchmark for Open Information Extraction
- WiRe57: A Fine-Grained Benchmark for Open Information Extraction
- CaRB: A Crowdsourced Benchmark for Open IE
- TextRunner: Open Information Extraction on the Web
- Span Model for Open Information Extraction on Accurate Corpus









#### Q/A Part



