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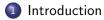
Fine-Grained Named Entities for Corona News

Sefika Efeoglu and Adrian Paschke

Freie Universität Berlin

February 15, 2023

Outline		





3 Evaluation





Introduction ••		

Motivation



Figure 1: The word cloud of corona news corpus from tagesschau.

Introduction	Methodology	Evaluation	Conclusion	Bibliography
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Introduction				

Problem:

- Huge amount of unstructured text data in the corona domain since December 2019.
- Analyzing these unstructured texts is time-consuming.

Drawbacks of existing corpora:

The existing corpora, such as CORD-19 [1] and LitCovid [2]:

- fail to identify recent variants of the coronavirus and generic mentions.
- include earlier published scientific papers in this domain.

Approach:

This study aims to develop an annotation pipeline that generates annotated training data from newer corona news articles for named entity recognition (NER).

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Methodology 0000		

An Annotation Pipeline for Training Texts

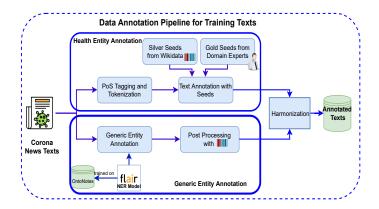
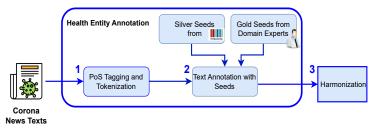


Figure 2: Data Annotation Pipeline for Training Texts.

Methodology 0000		

Health Entity Annotation



Input Sentence:

According to the Berlin virologist Christian Drosten, an unvaccinated person with an Omicron infection carries three quarters of the risk of being hospitalized for an unvaccinated person with the delta variant of Corona.

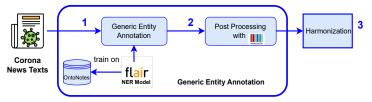
Output (Input to Step 3):

According to the Berlin virologist Christian Drosten, an unvaccinated person with an Omicron[CORONAVIRUS] infection[DISEASE_OR_SYNDROME] carries three quarters of the risk of being hospitalized for an unvaccinated person with the delta variant[CORONAVIRUS] of Corona[CORONAVIRUS].

Figure 3: Health Entity Annotation.

Methodology 0000		

Generic Entity Annotation



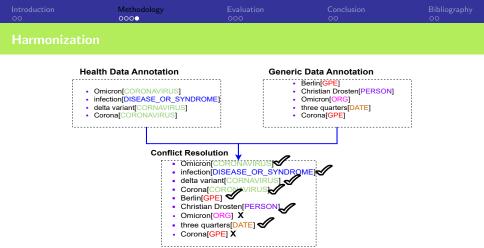
Input Sentence:

According to the Berlin virologist Christian Drosten, an unvaccinated person with an Omicron infection carries three quarters of the risk of being hospitalized for an unvaccinated person with the delta variant of Corona.

Output (Input to Step 3):

According to the <u>Berlin[OPE]</u> virologist <u>Christian Drosten[PERSON]</u>, an unvaccinated person with an <u>Omicron[ORG]</u> infection carries three quarters[DATE] of the risk of being hospitalized for an unvaccinated person with the delta variant of <u>Corona[GPE]</u>.

Figure 4: Generic Entity Annotation.



The Annotated Sentence:

According to the Berlin[GPE] virologist Christian Drosten[PERSON], an unvaccinated person with an Omicron[CORONAVIRUS] Infection[DISEASE_OR_SYNDROME] carries three quarters[DATE] of the risk of being hospitalized for an unvaccinated person with the delta variant[CORONAVIRUS] of Corona[CORONAVIRUS].

Figure 5: Harmonization

		Evaluation 000	
Experimental	Setup		

- **Dataset:** corona-related news articles from a German news-channel "Tagesschau" between December 2020 and June 2022.
- Fleiss Kappa: 0.98 (test) (calculated for event, product, immune_response, coronavirus, disease_or_sydrome, sign_or_symptom, 'empty').
- NER models: base (Glove) [3], advanced (Flair+Glove) [4] and SciBERT [5].

Corpus	\mid # of sentences
Training	89986
Dev	4999
Test	1000

Table 1: The entities in the data set have been categorized with 23 entity types.

	Evaluation 000	
Results		

- Fine-tuned SciBERT [5] model's micro F1-score is 0.7765.
- Its entity-specific F1-scores are 0.81 (coronavirus), 0.84 (sign_or_symptom), 0.79 (disease_or_syndrome), 0.8 (immune_response), and 0.85 (group).

Embedding	Model	Model Std	Coronavirus	Disease or Syndrome	Group	Imm ^{une} Response	sign or Symptom
Glove	0.71084	0.003414	0.76522	0.84152	0.80078	0.96364	0.81922
Glove+Flair	0.77162	0.002322	0.78614	0.81214	0.85016	0.83264	0.86562

Table 2: This table shows the statistical details about mean micro-F1 scores of the NER models (implemented by using Flair framework [6]), which were trained and evaluated five times. Besides, the table gives the mean micro-F1 scores of new entity types on the models trained with our corona news corpus.

	Evaluation 000	

F1 Scores of Specific Entities

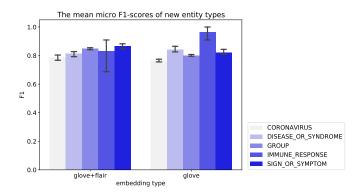


Figure 6: F1 scores of the new entities.

		Conclusion OO	
Conclusion			

- Contributions:
 - An annotation pipeline to create annotated texts from the corona news articles for NER.
 - A new up-to-date annotated corpus in the corona domain to identify corona-related mentions on the corona news articles via the NER models.
- The models utilizing contextual embedding surpass the model using an only word embedding in terms of micro-F1 score.
- Besides, the fine-tuned SciBERT model has performed well in the domain-specific entity types.

		Conclusion 00	Bibliography 00
Acknowledge	oments		

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	Methodology		Conclusion	Bibliography
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Thank you! sefika.efeoglu@fu-berlin.de

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