Improving scientific claim verification with weak supervision and full-document context

CS431.012.KHCL

Instructor: Nguyen Duy Khanh

Group 13

Le Gia Khang Nguyen Hoang Tan Le Duy Khang

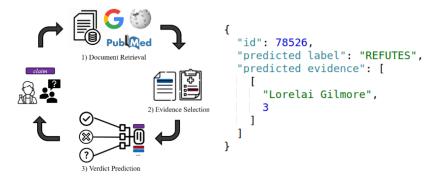
December 30, 2023

## Definition of scientific claim verification from the SCIFACT task:

Given a claim c and a candidate abstract a.

Label  $y(c, a) \in \{SUPPORTS, REFUTES, NEI\}.$ 

Identify rationales  $R(c, a) = \{r_1(c, a), \dots, r_n(c, a)\}.$ 



#### Claim:

Advil (ibuprofen) worsens COVID-19 symptoms

#### **Evidence abstract:**

# Covid-19 and avoiding Ibuprofen.

...

Increased risk of COVID-19 infection was feared with ibuprofen use

•••

#### Claim:

Advil (ibuprofen) worsens COVID-19 symptoms

Label: Refuted

#### **Evidence abstract:**

# Covid-19 and avoiding Ibuprofen.

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# Task Outputs

Fact-checking label

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# Task Outputs

- Fact-checking label
- Rationales justifying the label

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#### Label: Refuted

# Task Outputs

- Fact-checking label
- Rationales justifying the label

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# Covid-19 and avoiding Ibuprofen.

...

Increased risk of COVID-19 infection was feared with ibuprofen use Context required

..

#### Claim:

Advil (ibuprofen) worsens COVID-19 symptoms

#### **Evidence abstract:**

# Covid-19 and avoiding Ibuprofen.

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Increased risk of COVID-19 infection was feared with ibuprofen use

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At this time, there are no findings discouraging the use of ibuprofen

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Label: Refuted

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#### Claim:

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**Label: Refuted** 

#### **Evidence abstract:**

# Covid-19 and avoiding Ibuprofen.

Increased risk of COVID-19 infection was feared with ibuprofen use

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At this time, there are no findings discouraging the use of ibuprofen

#### Drawbacks of extract-then-label:

- Rationales may lack context
- Requires rationale supervision during training

# Prior models

Given a claim c and candidate abstract a

## These models make predictions in 2 steps:

Predict rationales  $\hat{R}(c, a) = \{\hat{r}_1(c, a), ... \hat{r}_n(c, a)\}$ 

Then, make a label prediction  $\hat{y}(c, f_R(\hat{R}(c, a)))$ 

## A multitask system for full-context scientific claim verification

- Predict  $\hat{y}(c, a)$  directly based on an encoding of the entire claim and abstract.
- Enforce consistency of  $\hat{R}(c, a)$  with  $\hat{y}(c, a)$  during decoding.

### Long document encoding:

A claim c and candidate abstract a consisting of title t and sentences

$$s_1, ..., s_n$$
.

The </s> token following each sentence  $s_i$  is notated as  $</s>_i$ .

$$~~c~~ts_1_1 ... s_n_n$$

Global attention is assigned to < s > token, all tokens in c and all < /s > tokens.

# $\mathsf{MultiVerS}$

#### Claim

Advil (ibuprofen) worsens COVID-19 symptoms

#### Abstract sentence 1

Increased risk of COVID-19 infection was feared with ibuprofen use

#### Abstract sentence N

#### Globally-contextualized representation using Longformer

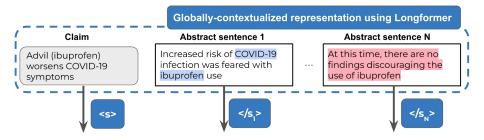
#### Claim

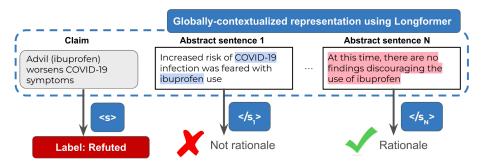
Advil (ibuprofen) worsens COVID-19 symptoms

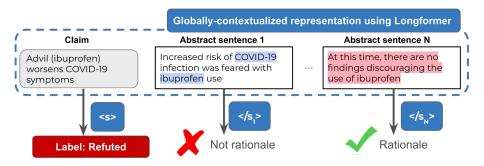
#### Abstract sentence 1

Increased risk of COVID-19 infection was feared with ibuprofen use

#### Abstract sentence N







$$\mathcal{L} = \mathcal{L}_{ extit{label}} + \lambda_{ extit{rationale}} \mathcal{L}_{ extit{rationale}}$$

### Benefits of multitask approach:

- Incorporates all relevant context
- 2 Can train on instances with no rationale annotations

Dataset	Domain	Claim source	Open	Has NEI	Claim complexity	Negation method	Train claims	Eval claims	> 512 tokens	
HealthVer	COVID	TREC-COVID	×	1	Complex	Natural	1,622	230	14.9%	
COVIDFact	COVID	Reddit	X	×	Complex	Automatic	903	313	12.4%	
SCIFACT	Biomed	Citations	✓	✓	Atomic	Human	1,109	300	27.4%	
FEVER	Wiki	Wikipedia	-	/	Atomic	Human	130,644	-	33.2%	
PUBMEDQA	Biomed	Paper titles	-	✓	Complex	Automatic	58,370	-	12.1%	
EVIDENCEINFERENCE	Biomed	ICO prompts	-	/	Atomic	Automatic	7,395	-	42.7%	

Table: Summary of datasets used in experiments

### Target datasets:

- HealthVer
- COVID-Fact
- SciFact

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- HealthVer
- COVID-Fact
- SciFact

Roughly 1000 claims / dataset.

Expert annotations are expensive

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HealthVer

Roughly 1000 claims / dataset.

COVID-Fact

Expert annotations are expensive

SciFact

### Traning procedure:

- **Stage 1:** Train on a combination of *labeled out of domain* data weakly-labeled in-domain data.
- Stage 2: Continue training on data from each target dataset.

### Target datasets:

HealthVer

Roughly 1000 claims / dataset.

COVID-Fact

Expert annotations are expensive

SciFact

### Traning procedure:

- **Stage 1:** Train on a combination of *labeled out of domain* data weakly-labeled in-domain data.
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## Domain adaptation settings:

- **Zero-shot:** Stage 1 training only.
- Few-shot: 45 instances from target datasets.
- Full-supervised: All target data.

**Supervised** out-of-domain data (FEVER)

LeBron James was born in

Ohio

Label: Supported

LeBron James is an American basketball player. He was born in Akron, Ohio.

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LeBron James is an American basketball player. He was born in Akron, Ohio. Weakly-supervised in-domain data

Diabetes increases risk of depression

Abstract

...

Label: Supported

**Supervised** out-of-domain data (FEVER)

LeBron James was born in

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Label: Supported

LeBron James is an American basketball player. He was born in Akron, Ohio. Weakly-supervised in-domain data

Diabetes increases risk of depression Abstract ...

Label: Supported

Claim: Paper title

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Diabetes increases risk of depression

Abstract
...
Rati apprare
...
Label: Supported

Claim: Paper title

Rationales likely to appear in abstract, but are not annotated

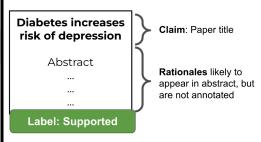
**Supervised** out-of-domain data (FEVER)

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Label: Supported

LeBron James is an American basketball player. He was born in Akron, Ohio. Weakly-supervised in-domain data



MultiVerS can train on these examples, even though no rationale annotations are provided

## **Evaluation**

#### Abstract-level evaluation:

- Identifying abstracts that SUPPORT or REFUTE each claim.
- Predicting the correct label y(c, a) is sufficient

#### Sentence-level evaluation:

• It combines the accuracy of abstract-level label prediction with the precision of rationale identification.

# **Evaluation**

	Model	HealthVer				COVIDFact						SCIFACT							
Setting		Abstract		Sentence		Abstract			Sentence			Abstract			Sentence				
		P	R	F1	P	R	F1	P	R	F1	P	R	F1	P	R	F1	P	R	F1
Zero	PARAGRAPHJOINT	72.3	14.4	24.0	22.9	2.7	4.9	51.3	37.9	43.6	31.5	16.0	21.3	52.9	32.4	40.2	36.4	14.9	21.1
	MULTIVERS	60.6	20.5	30.7	25.0	4.6	7.8	48.8	45.7	47.2	32.7	18.5	23.6	49.0	44.6	46.7	39.0	21.6	27.8
Few	PARAGRAPHJOINT	62.7	41.6	50.0	46.0	29.3	35.8	73.3	60.6	66.3	44.3	30.6	36.2	44.4	51.4	47.6	33.0	35.1	34.0
	MULTIVERS	63.6	47.9	54.7	41.9	31.0	35.7	71.3	68.1	69.7	39.5	35.4	37.4	76.4	54.1	63.3	51.7	40.3	45.3
Full	VERT5ERINI PARAGRAPHJOINT			72.6 71.5									33.9 40.8			68.2 69.1			
	MULTIVERS	78.9	76.3	77.6	71.4	67.0	69.1	77.3	77.3	77.3	41.5	46.1	43.7	73.8	71.2	72.5	67.4	67.0	67.2

Table 2: Performance of MultiVerS and baselines.

### covidfact\_covidfact.jsonl:

```
"abstract_label_only_precision": 0.7728706624605678,
"abstract_label_only_recall": 0.7728706624605678,
"abstract_label_only_f1": 0.7728706624605678,
"sentence_label_precision": 0.41694915254237286,
"sentence_label_recall": 0.458955223880597,
"sentence_label_f1": 0.43694493783303723
```

### covidfact\_fever.jsonl:

```
"abstract_label_only_precision": 0.958333333333333334,
"abstract_label_only_recall": 0.14511041009463724,
"abstract_label_only_f1": 0.25205479452054796,
"sentence_label_precision": 0.631578947368421,
"sentence_label_recall": 0.04477611940298507,
"sentence_label_f1": 0.08362369337979093
```

### covidfact\_fever\_sci.jsonl:

```
"abstract_label_only_precision": 0.4882154882154882,
"abstract_label_only_recall": 0.45741324921135645,
"abstract_label_only_f1": 0.4723127035830619,
"sentence_label_recall": 0.0914179104477612,
"sentence_label_f1": 0.15654952076677317
```

### healthver\_fever.jsonl:

```
"abstract_label_only_precision": 0.8,
"abstract_label_only_recall": 0.00667779632721202,
"abstract_label_only_f1": 0.013245033112582783,
"sentence_label_precision": 1.0,
"sentence_label_recall": 0.0027223230490018148,
"sentence_label_f1": 0.005429864253393665
```

### healthver\_fever\_sci.jsonl:

```
"abstract_label_only_precision": 0.6059113300492611,
"abstract_label_only_recall": 0.20534223706176963,
"abstract_label_only_f1": 0.30673316708229426,
"sentence_label_precision": 0.40625,
"sentence_label_recall": 0.011796733212341199,
"sentence_label_f1": 0.022927689594356263
```

#### healthver\_healthver.jsonl:

```
"abstract_label_only_precision": 0.7892918825561313,
"abstract_label_only_recall": 0.7629382303839732,
"abstract_label_only_f1": 0.7758913412563666,
"sentence_label_precision": 0.7194525904203324,
"sentence_label_recall": 0.6678765880217786,
"sentence_label_f1": 0.6927058823529412
```

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#### scifact\_fever.jsonl:

```
"abstract_label_only_precision": 0.8378378378378378,
"abstract_label_only_recall": 0.13963963963963963,
"abstract_label_only_f1": 0.23938223938223935,
"sentence_label_precision": 0.7307692307692307,
"sentence_label_recall": 0.051351351351351354,
"sentence_label_f1": 0.09595959595959597
```

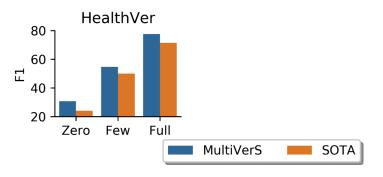
#### scifact\_fever\_sci.jsonl:

```
"abstract_label_only_precision": 0.4900990099009901,
"abstract_label_only_recall": 0.44594594594594594,
"abstract_label_only_f1": 0.4669811320754717,
"sentence_label_precision": 0.7017543859649122,
"sentence_label_recall": 0.10810810810810811,
"sentence_label_f1": 0.1873536299765808
```

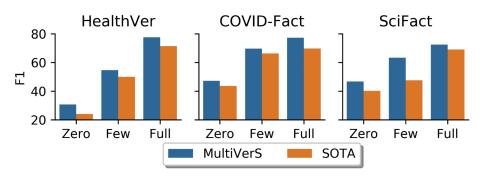
#### scifact\_scifact.jsonl:

```
"abstract_label_only_precision": 0.7383177570093458,
"abstract_label_only_recall": 0.7117117117117117,
"abstract_label_only_f1": 0.7247706422018348,
"sentence_label_precision": 0.6739130434782609,
"sentence_label_recall": 0.6702702702702703,
"sentence_label_f1": 0.6720867208672086
```

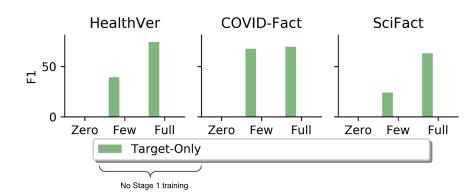
## Results

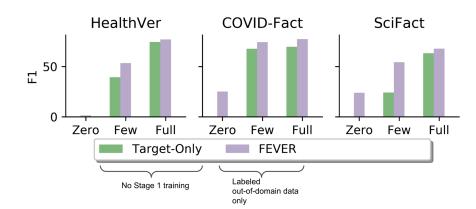


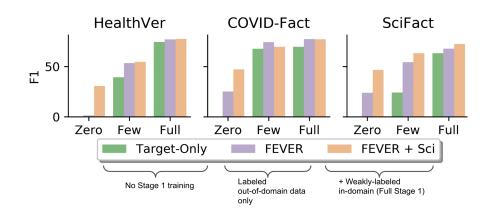
### Results

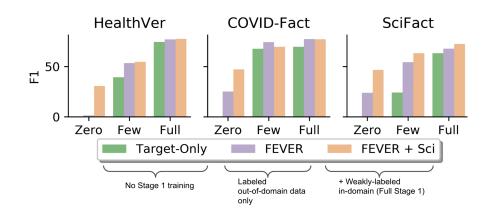


MultiVerS outperforms SOTA on all datasets









Pretraining with weakly-supervised in-domain data improves few / zero shot performance.

### Reference



MULTIVERS: Improving scientific claim verification with weak supervision and full-document context



Scientific Fact-Checking: A Survey of Resources and Approaches Jurai Vladika and Florian Matthes



Longformer: The Long-Document Transformer

Iz Beltagy, Matthew E. Peters and Arman Cohan



Code and model checkpoints for the MultiVerS model

dwadden/multivers

# Thanks for listening!

Q&A section