



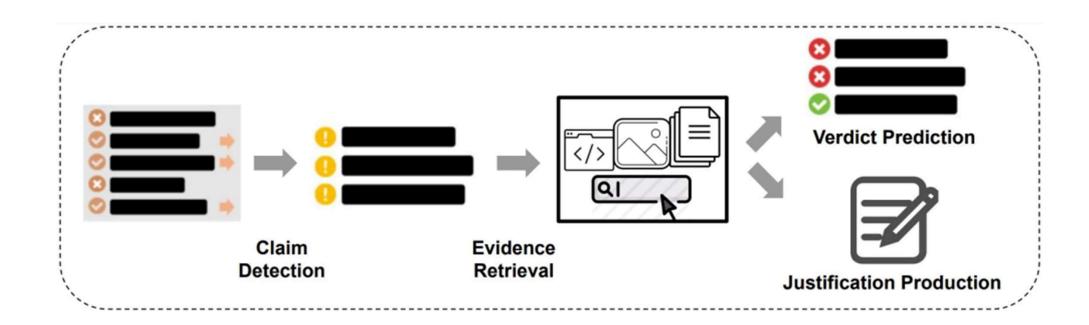


CHEF: A Pilot Chinese Dataset for Evidence-Based Fact-Checking

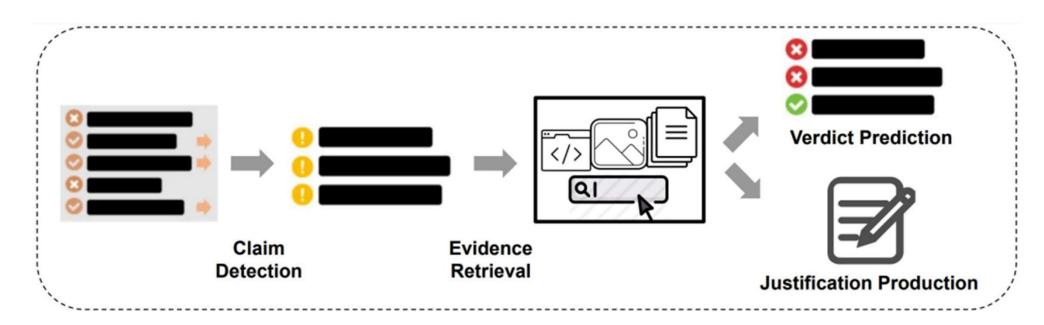
Xuming Hu^{1*}, Zhijiang Guo^{2*}, Guanyu Wu¹, Aiwei Liu¹, Lijie Wen¹, Philip S. Yu^{1,3}

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Automated Fact Checking



Automated Fact Checking



- A handful of non-English Datasets.
- Claims are created by non-English articles.

Dataset Comparisons

Dataset	Natural	Domain	#Claims	Languaga		Evi	dence	
Dataset	Naturai	Domain	#Claims	Language	Type	Source	Retrieved	Annotaated
FEVER (Thorne et al., 2018)	×	Multiple	185,445	English	Text	Wiki	/	1
HOVER (Jiang et al., 2020)	×	Multiple	26,171	English	Text	Wiki	1	1
TabFact (Chen et al., 2020)	X	Multiple	92,283	English	Table	Wiki	X	✓
InfoTabs (Gupta et al., 2020)	X	Multiple	23,738	English	Table	Wiki	×	1
ANT (Khouja, 2020)	×	Multiple	4,547	Arabic	×	X	×	X
VitaminC (Schuster et al., 2021)	×	Multiple	488,904	English	Text	Wiki	×	✓
DanFEVER (Nørregaard and Derczynski, 2021)	X	Multiple	6,407	Danish	Text	Wiki	✓	1
FEVEROUS (Aly et al., 2021)	×	Multiple	87,026	English	Text/Table	Wiki	1	1
PolitiFact (Vlachos and Riedel, 2014)	1	Politics	106	English	Meta/Text	FC	×	X
PunditFact (Rashkin et al., 2017)	1	Multiple	4,361	English	×	X	×	X
Liar (Wang, 2017)	1	Multiple	12,836	English	Meta	FC	X	X
Verify (Baly et al., 2018)	1	Politics	422	Mul(2)	Text	Internet	1	X
MultiFC (Augenstein et al., 2019)	1	Multiple	36,534	English	Meta/Text	Internet	1	X
Snopes (Hanselowski et al., 2019)	1	Multiple	6,422	English	Text	FC	×	X
SciFact (Wadden et al., 2020)	1	Science	1,409	English	Text	Paper	×	X
PUBHEALTH (Kotonya and Toni, 2020b)	1	Health	11,832	English	Text	FC	×	X
AnswerFact (Zhang et al., 2020)	1	Product	60,864	English	Meta/Text	Amazon	1	X
FakeCovid (Shahi and Nandini, 2020)	1	Health	5,182	Mul(3)	×	X	X	X
XFact (Gupta and Srikumar, 2021)	1	Multiple	31,189	Mul(25)	Meta/Text	Internet	1	×
CHEF	1	Multiple	10,000	Chinese	Meta/Text	Internet	1	1

- Natural
- Synthetic

Table: Comparisons of fact-checking datasets.

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Synthetic:

- Restricted world knowledge to a single source.
- Claims created artificially by mutating sentences from Wikipedia articles.

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Natural:

- Fact checking websites are small in size.
- Summary
 snippets do not
 provide
 sufficient
 information.

Table: Comparisons of fact-checking datasets.

CHEF

Claim: 2019年, 共有12.08万人参加成都中考,但招生计划只有4.3万。 In 2019, a total of 120,800 students participated in the high school entrance examination in Chengdu, but schools only enrolled 43,000 students.

Document: 今年共有12.08万人参加中考,这个是成都全市,包括了20个区,高新区和天府新区的总参考人数。 月前,教育局公布了2019年的普高招生计划。招生计划数进一步增加,上普高的机会更大了... 中心城区(13个区)招生计划为43015人。 This year, 120,800 people participated in the high school entrance examination. This number is for the entire city of Chengdu, including 20 districts, high-tech zone and Tianfu new district. A month ago, the Education Bureau announced the 2019 high school enrollment plan. The number of enrollment will be increased, indicating that there is a greater chance of going to high school... The plan of the central area (including 13 districts) is 43,015.

Verdict: Refuted; Domain: Society

Challenges: Evidence Collection; Numerical Reasoning

Table: An example from CHEF.

CHEF: CHinese dataset for Evidence-based Fact-checking

- 10,000 real-world claims
- 6 Chinese fact-checking websites
- Annotated evidence
- Developed suitable guidelines
- Performed data validation

- Data collection
- Claim labeling
- Evidence retrieval
- Data validation

• Data collection

Website	Domain	URL	Total
Piyao TFC Mygopen Jiaozhen	Multiple Multiple Multiple Multiple	www.piyao.org.cn tfc-taiwan.org.tw www.mygopen.com vp.fact.qq.com	3,741 1,759 1,654 157
Cnews	Multiple	m.chinanews.com	2,689
Total	Multiple	-	10,000

Table: Statistics of data source.

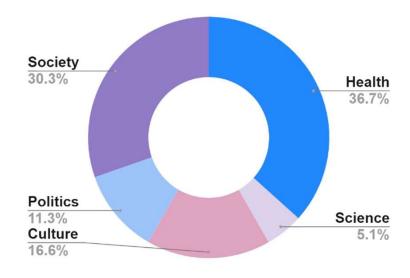


Figure: Distributions of domains.

Claim Labeling

Split	SUP	REF	NEI	Total			
Train	2,877	4,399	776	8,002			
Dev	333	333	333	999			
Test	333	333	333	999			
Avg #V	Words in	the Clain	1	28			
Avg #V	Words in	the Goog	le Snippets	68			
Avg #V	Avg #Words in the Evidence Sentences 126						
Avg #V	Words in	the Source	ce Documents	3,691			

Table: Dataset split sizes and statistics for CHEF.

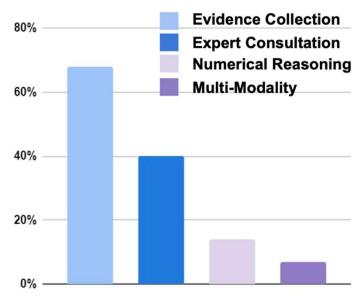
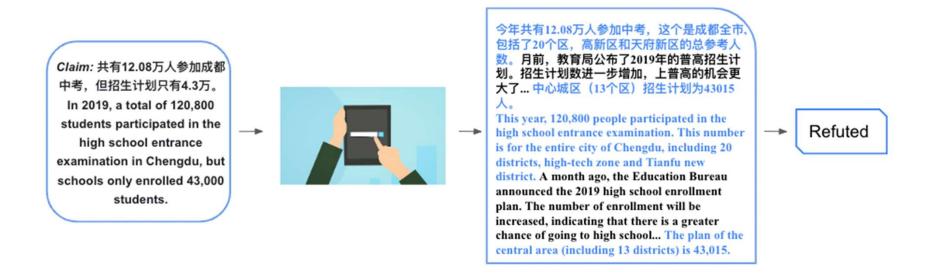


Figure: Distributions of challenges.

Evidence Retrieval



Source Documents

The claim is refuted by the evidence, which are sentences retrieved (highlighted) from the document.

Data Validation

5-way inter-annotator agreement

- 310 Claims
- 5 Annotators

Fleiss K score = 0.74

Another 310 Claims

- 88.7% were labeled correctly
- 83.6% provided sufficient information

Pipeline Systems

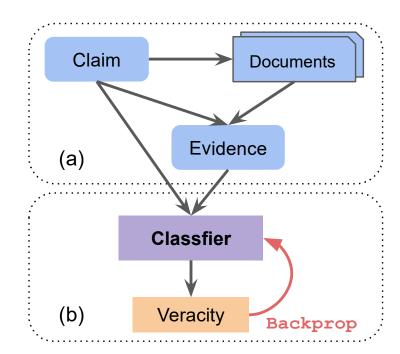


Figure: Pipeline Systems

Pipeline Systems

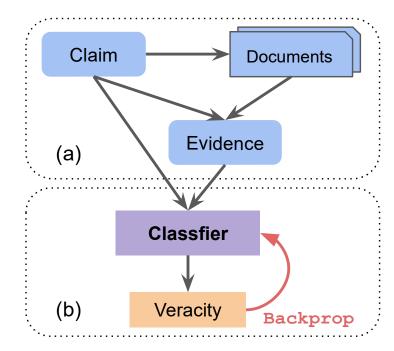


Figure: Pipeline Systems

Evidence Retrieval

- Surface Ranker: TF-IDF
- Semantic Ranker: Cosine similarity
- Hybrid Ranker: RankSVM
- Google Snippets: Google Search Engine

• Pipeline Systems

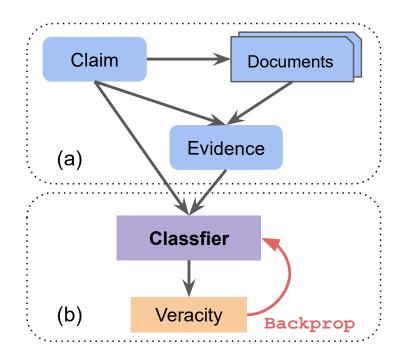


Figure: Pipeline Systems

Veracity Prediction

- BERT-Based Model
- Attention-Based Model
- Graph-Based Model

Joint Systems

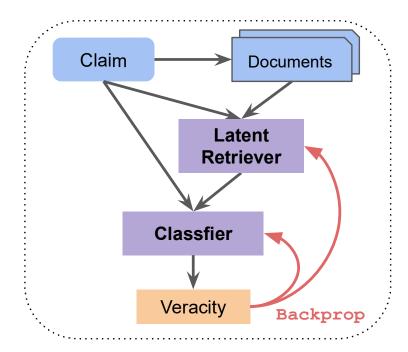


Figure: Joint Systems

Latent Retriever

Hard Kumaraswamy distribution (Bastings et al., 2019)

- More Baselines
- Reinforce (Lei et al. 2016)
- Multi-task (Yin and Roth 2018)

Main Results

9	System / Evidence			BERT-Based Model ¹		ased Model ²	Graph-Based Model ³	
				Macro F1	Micro F1	Macro F1	Micro F1	Macro F1
	No Evidence		54.46±2.89	52.49±2.44	54.87±1.95	53.47±2.82	_	_
	Snippets	ľ	62.07 ± 2.55	60.61±2.96	62.42±2.31	60.24±2.56	62.78 ± 1.70	61.06±2.59
D:1:	Surface Rank	ker	63.17±1.67	61.47±2.02	63.77±1.89	62.65±2.32	64.58±1.45	61.46±1.72
Pipeline	Semantic Ra	Semantic Ranker		61.94 ± 1.66	63.95 ± 1.46	62.80 ± 1.33	64.67 ± 1.54	62.28 ± 1.50
	Hybrid Ranker		63.29 ± 1.65	61.80 ± 2.31	63.48 ± 1.22	62.74 ± 1.30	64.37 ± 1.66	62.58 ± 1.43
	Reinforce ⁴	Snippets	63.76±1.52	61.74±1.88	64.06±1.76	61.97±1.04	65.77±1.23	62.34±1.11
		Documents	64.37±1.65	62.46±1.72	64.86±1.83	62.66±1.32	66.58±1.45	63.47±1.58
Joint	N. 10 . 15	Snippets	62.78 ± 1.41	61.98±2.59	64.43±1.72	61.58±1.34	66.21±1.57	63.15±1.46
	Multi-task ⁵	Documents	65.02±1.46	63.12±1.78	65.45±1.59	62.94 ± 2.03	67.46±1.72	64.31±1.81
		Snippets	64.45±1.68	62.52 ± 2.23	65.73±1.75	63.44±1.68	67.81±1.74	64.34±1.57
	Latent	Documents	66.77±1.43	64.65±1.74	67.62±1.48	64.81±1.26	69.12±1.13	65.26±1.67
Pipeline	Pipeline Gold Evidence		78.99±0.82	77.62±1.02	79.18±1.07	78.36±1.40	79.84±1.24	78.47±1.17

1. Evidence plays an important role in verifying realworld claims.

Schuster et al. (2021)¹, Gupta and Srikumar (2021)², Liu et al. (2020)³, Lei et al. (2016)⁴, Yin and Roth (2018)⁵

Table: Main results.

Main Results

	System / Evidence		BERT-Bas	ed Model ¹	Attention-B	ased Model ²	Graph-Based Model ³	
			Micro F1	Macro F1	Micro F1	Macro F1	Micro F1	Macro F1
	No Evidence		54.46±2.89	52.49±2.44	54.87±1.95	53.47±2.82	_	_
	Snippets		62.07 ± 2.55	60.61 ± 2.96	62.42 ± 2.31	60.24 ± 2.56	62.78 ± 1.70	61.06 ± 2.59
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Pipeline	Semantic Ranker Hybrid Ranker		63.47 ± 1.71	61.94 ± 1.66	63.95 ± 1.46	62.80 ± 1.33	64.67 ± 1.54	62.28 ± 1.50
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Table: Main results.

- 1. Evidence plays an important role in verifying real-world claims.
- 2. Retrieving evidence sentences from documents achieve better F1 scores than directly use the summary snippets.

Main Results

	System / Evidence		BERT-Bas	ed Model ¹	Attention-Based Model ²		Graph-Based Model ³		
•			Micro F1	Macro F1	Micro F1	Macro F1	Micro F1	Macro F1	
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Table: Main results.

- Evidence plays an important role in verifying real-world claims.
- Retrieving evidence sentences from documents achieve better F1 scores than directly use the summary snippets.
- Joint system outperforms pipeline system consistently with both Google snippets and source documents as inputs.

• Effect of Evidence

#E	GS	Sur	Sem	Hyb	JG	JS
1	55.24	55.67	56.04	56.72	56.98	57.54
					59.89	
5	60.61	61.47	61.94	61.80	62.12	64.65
10	59.12	60.20	60.37	61.24	61.86	64.73
15	55.72	56.31	56.56	57.08	58.69	59.11

Table: Effect of Evidence. #E indicates the number of evidence.

The fluctuation results indicate that both **quantity and quality** of retrieved evidence affect the performance.

- Fewer evidence -> incomplete coverage
- More evidence -> irrelevant sentences

Performance against Claim Length

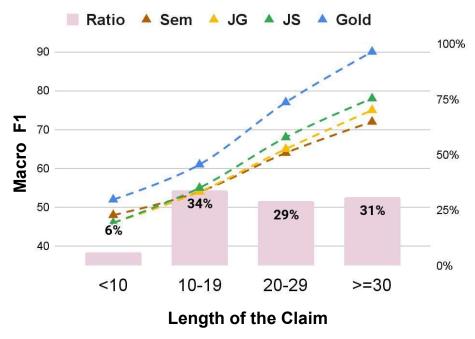
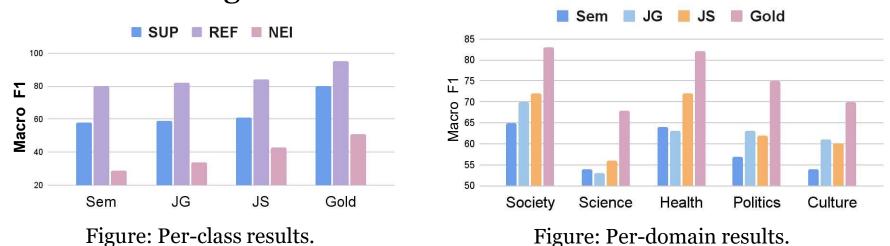


Figure: Comparisons against claim lengths.

- 1. Most claims are longer than **10** words.
- 2. Performance of the systems on short claims is lower than other.

Performance against Classes and Domains



- 1. The scores of minor classes are much lower than the majority class.
- 2. Claims from science, politics and culture domains have fewer training instances as most claims in the dataset focus on the society and public health topics.

THANK YOU!



Code + Data are Available at:

http://github.com/THU-BPM/CHEF hxm19@mails.tsinghua.edu.cn