NumClaim: Investor's Fine-grained Claim Detection



Chung-Chi Chen¹, Hen-Hsen Huang^{2,3}, Hsin-Hsi Chen^{1,3}

¹Department of Computer Science and Information Engineering, National Taiwan University, Taiwan ²Department of Computer Science, National Chengchi University, Taiwan ³MOST Joint Research Center for AI Technology and All Vista Healthcare, Taiwan

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Overview



- Argument mining issue in finance
- Expert-annotated dataset, NumClaim
- We show that encoding with numeral encoder and cotraining with the numeral understanding auxiliary task are helpful for the numeral-oriented task.

Motivation



- Over 58.47% of sentences in analysis report contain at least one numeral
- Investors always make a claim with an estimation
 - (X) We estimate that the sales may growth
 - (O) We estimate that the sales growth rate may exceed 40%
- The importance of fine-grained claims and the numerals.
 - We estimate that the sales growth rate may exceed 5%
 - We estimate that the sales growth rate may exceed 40%



NumClaim



- Chinese financial analysis reports
- The annotators work in the financial industry (bank's treasury department and hedge fund)
- The Cohen's kappa agreements between the experts are 88.31%
- 5,144 instances: 23.78% "In-claim" and 76.22% "Out-of-claim"

Sentence	Label
We estimate that the sales growth rate may exceed 40 %.	In-claim
Professional audio/visual products account for 20%.	Out-of-claim

In-claim	1	Out-of-claim	
estimate	2.86	lower/higher than	-1.37
price target	2.80	cause	-1.37
downgrade	2.58	last year	-1.26
upgrade	1.55	influence	-1.25



Auxiliary Task – Numeral Understanding



• The Cohen's kappa agreements between the experts are 89.55%

Category	Subcategory	In-claim	Out-of-claim	Sum
	price	42	33	75
Monetary	money	506	368	874
	change	3	15	18
Porcontago	absolute	208	500	708
Percentage	relative	408	402	810
Tomporal	date	0	2,134	2,134
Temporal	time	0	3	3
Quantity	absolute	55	219	274
	relative	0	4	4
Product Nur	nber	1	135	136
Ranking		0	3	3
Other		0	105	105
Te	otal	1,223	3,921	5,144



Chung-Chi Chen, Hen-Hsen Huang, Yow-Ting Shiue, and Hsin-Hsi Chen. 2018. Numeral understanding in financial tweets for fine-grained crowd-based forecasting. In *IEEE/WIC/ACM International Conference on Web Intelligence*

Statistics



Dataset	NumClaim	CRC [13]	PE [12]
Language	Chinese	Chinese	English
Source	Analysis Report	Hotel Review	Persuasive Essay
# Word	42,594	21,848	97,420
# Numeral	5,144	67	111

	NumClaim	CRC
# hard words	31.95	18.28
# negative words	0.14	0.60
# synonym	0.28	1.49
Noun phrase modifier ratio	0.29	0.38
Noun phrase ratio	31.79	26.62
# transition words	4.86	1.62

[12] Steffen Eger, Johannes Daxenberger, and Iryna Gurevych. 2017. Neural End-to-End Learning for Computational Argumentation Mining. In ACL

[13] Steffen Eger, Johannes Daxenberger, Christian Stab, and Iryna Gurevych. 2018. Cross-lingual Argumentation Mining: Machine Translation (and a bit of Projection) is All You Need!. In COLING.



Experimental Results



- Encoding: BERT
- Baseline: CNN, BiGRU, CapsNet
- Metrics: Macro-F1
- Class Weight (CW)
- Numeral Encoder
 - Represent the digit (0--9) and the decimal point as a 11dimension tensor, and concatenate it with a tensor for the inter-numeral position information.
- Joint Learning with Category Classification Task (CG)

Architecture	CNN	BiGRU	CapsNet
Baseline	76.15%	77.97%	77.93%
+ CW	77.26%	78.29%	78.68%
+ CW & NE (CNN)	78.19%	79.06%	80.91%
+ CW & NE (CNN) & CG	81.35%	81.65%	82.62%



Conclusion & Future Direction



- Our contributions
 - Explore the argument mining issue in finance
 - Provide an expert-annotated dataset NumClaim
 - Propose helpful methods for solving numeral-oriented task
- Future Directions Fine-grained Financial Opinion Mining
 - Premise detection and relation linking
 - Rationality assessment



Chung-Chi Chen, Hen-Hsen Huang, and Hsin-Hsi Chen. 2020. Fine-grained Financial Opinion Mining: A Survey and Research Agenda. In *arXiv:2005.01897*



Related Datasets and Events



- FinNum-1: Fine-Grained Numeral Understanding in Financial Tweets (NTCIR-14, 2018-2019)
- FinNum-2: Numeral Attachment in Financial Tweets (NTCIR-15, 2019-2020)
- FinNum-3: Investor's Fine-grained Argument Detection (Will submit proposal to NTCIR-16)
- Tutorial in AACL-IJCNLP 2020: Natural Language Processing in Financial Technology Applications
- Springer SpringerBriefs: Financial Opinion Mining (Available in 2021)

Chung-Chi Chen, Hen-Hsen Huang, and Hsin-Hsi Chen. 2020. NLP in FinTech Applications: Past, Present and Future. In arXiv:2005.01320.

Feel free to contact us if you have any questions.

Chung-Chi Chen: cjchen@nlg.csie.ntu.edu.tw



