TMCIT at the NTCIR-14 QALab-PoliInfo Task

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Motivation

In recent years, a lot of fake information exists on the Internet. In particular, political fake information has a large impact on society.

> It is necessary to **argue based on the evidence** to avoid being misled.

Purpose

A system that presents information necessary for discussion.

Classify the assembly member speeches

Classification experiment results

	Acc	Prec[%]			Rec[%]			
	[%]	0	1	2	0	1	2	
Relevance	89	38	99	-	77	83	-	
Fact-checkability	80	92	63	-	68	71	-	
Stance	84	91	44	41	86	46	14	
Final Class	93	99	17	15	93	35	6	

Meaning of label

- Relevance
 - 0: irrelevant
 - 1: relevant
- Fact-checkability
 0: impossible
 1: possible
- Final Class

that are the source of inf	ormation.
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Final Class \ Class	Relevance	Fact-checkability	Stance	
Fact-checkable Agreement	relevant	possible	agree	
Fact-checkable disagreement	relevant	possible	disagree	
Other	irrelevant	impossible	neutral	



- Stance
 - 0: neutral
 - 1: agreement
- 0: other
- 1: Fact-checkable agreement
- 2: Fact-checkable disagreement

Discussion

Classification of Relevance

- 2: disagreement

- Label integration was made so that there was a lot of irrelevance.
- Low Kappa value for labels given by multiple people.

\rightarrow Decrease in precision rate.

- Top 10 most important words for relevance.
 - (Should promote integrated resorts including casinos.)
 - irrelevance: ten, hundred, chairman, two, six, eight, seven,
 Daio Paper, illegality,10 billion yen.
 - relevance: Casino, things, ir, attract, of, bringing in, to, integrated resort, country, facility.
 - \rightarrow tf-idf has successfully extracted high importance words.

Classification of Fact-checkability

- Among the nodes of the constructed decision tree, observe the node that are the number of samples ≥ 50 and impurity(gini) ≤ 0.1.
 - <u>influential</u>: Number of words representing Numeral / Time / Money
 / Country / Place and NER.

Classification of Fact-checkability by Decision tree

- Build a decision tree with manually labeled data.
- Use number of expressions as feature.
 - Evidence: "for"(ため), "so"(ので), "because"(から), "therefore"(したがって)
 - Numeral/Time/Money/Percentage: hundred, Heisei, yen, %, etc.
 - Named entity such as person's name/area/organization: Koike,
 Chinese, Jiminto, etc.
- Categorize test data by the tree as fact-checkable or not.

Classification of Stance by Support Vector Machine

- Use the emotional polarity value of the words of each sentence for the feature.
- Principal vector analysis reduces feature vectors to 150 dimensions.

		反対	危険	賛成	可決	利点		
1. 私は反対(-0.88)です.	1	-0.88	0	0	0	0]	
2. 危険(-0.99)なので反対(-0.88)です.	2	-0.88	-0.99	0	0	0		
3. 私は替成(0.99)です.	3	0	0	0.99	0	0		

- <u>uninfluential</u>: Number of expressions of Evidence / Percentage / Percentage name / organization.
- The reason why "the number of evidence expressions" is uninfluential.
 - Unable to properly extract evidence expressions.
 - Ex.: kind of "for" (in Japanese "ため").
 - Evidence expressions showing **cause** and **reason**.
 - Objective expressions showing **profit** or **goal**.
 - \rightarrow Need to define more complex extraction rules.

Classification of Stance

- Low accuracy rate and recall rate for agreement and disagreement
 - $\circ~$ The percentage of labels is,
 - neutral : agreement : disagreement = 80 : 12 : 8.
 - \circ Feature is simple.
 - Polarity reversal or expression is not captured well.

 \rightarrow Needs to incorporate the appropriate knowledge for the domain.

Conclusion

- we classified the utterances of assemblymen according to three viewpoints: Relevance, Fact-checkability and Stance.
- In the minority class of each classification experiment, the scores of



precision and recall were low.

• Future work

Improve quality of use data.

More complex extraction rules and feature definitions.