November 13-16, 2020 Online Conference

FSDM 2020

Keyword Extraction from TV Program Viewers' Tweet based on Neural Embedding Model



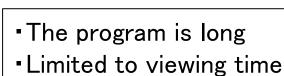
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Introduction

Problem

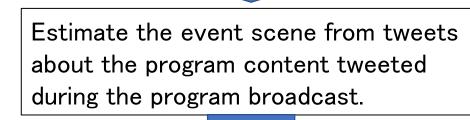
The number of people who watch television (TV) programs by recording them has increased in recent years.



Difficult to watch the entire program

The solution shown in this study

More viewers tweet while watching TV programs on Twitter.



Reduce viewing time

3. Data

- Collect tweet data of 3 programs of "Variety", "Movie" and "Anime".
- The collection time is from 10 minutes before the start of the program to 10 minutes after the end of the program.
- Prepared the correct answer data attached manually.

Table 1 Details of the collected programs

able 1. Details of the collected programs				Table 2. Number of scenes in each program category				
roadcaster	On-air date	Program name	Number of tweets	Category	Yofukashi	Friday	Pokémon	Total
Nippon	2019/10/14 Getsuyou Yofuka	Getsuyou Kara	14239	Person	5	13	7	25
Television		Yotukashi		Action	8	70	4	82
Nippon Television	2019/11/8	Friday Road SHOW – It	19460	Emotion	1	1	1	3
				Summary	7	2	0	9
TV TOKYO	2019/ 12/8	Pokémon	4105	Scenery	2	19	2	23
				Conversion	26	17	9	52
				Total	49	122	23	194

2. Proposed Method

Tweet collection and vectorization

- Collect the tweet text containing "# [program name]" using the Twitter API
 - Divide the tweet set of each program into 1 minute units
- Vectorization using BERT, is trained model on the Japanese Wikipedia corpus

BERT

- BERT is a deep learning model for natural language processing that by Google in 2018.
- Compared to Word2vec, it has the feature that the entire sentence can be vectorized.

4. Result

Experimental method:

- Using the correct answer data as a search query, calculate the top 5 time zones based on the similarity with the key phrase.
- t-SNE is used to compress the dimensions and visualize them on the two-dimensional coordinate space.

Evaluation method:

- Compare the time of correct data and the error of output result
 - Calculate difference between the correct time and the predicted time
- Also, compare with MMR and without MMR

Table 3. Average error of each program

Raishu no Anipoke wa Kabigon kyodaika desu ! Morphological analysis with MeCab+Neologd Raishu / no / Anipoke / wa / Kabigon / kyodai / ka / desu / !

With the part-of-speech pattern of "proper noun + adjective" as the

condition, phrases that match it are candidates

Vector is calculated by calculating sentence embedding for candidate

phrases and entire sentences

Candidate phrases close to the vector of the document itself are

extracted based on the degree of similarity, and key phrases are

determined.

Anipoke, Kabigon

Key phrase extraction

Many combinations of "noun + adjective" can be seen in live tweets.



- The EmbedRank algorithm mainly extracts patterns of "proper nouns" + "adjectives"
- The EmbedRank algorithm extracts candidate phrases extracted based on part of speech patterns and important phrases based on sentence vectors.

Experimental result:

Event scene search results: Less error with MMR than without MMR.

With MMR Without MMR 9 min. 41 sec. 14 min. 43 sec. Yofukash 16 min. 3 sec. 11 min. 3 sec. **Friday** 3 min. 33 sec. 5 min. 38 sec. Pokémon

Scene classification and visualization

• Since many proper nouns and adjectives are

"action" scene classification.

included in the tweet, the result is that many

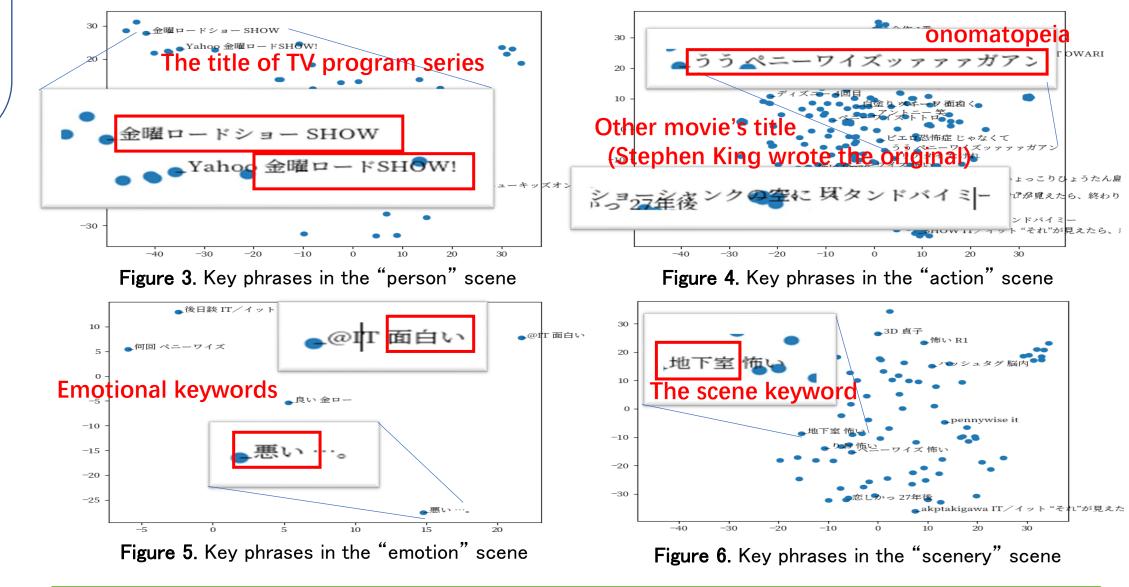
emotional words and nouns are included in the

Even in the "emotion" and "summary" scenes,

the verbs, adverbs, and general nouns that are

necessary to describe the scene could not be

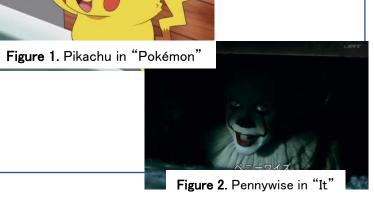
Key phrase visualization for each scene category in "Friday Road SHOW":



Key phrase extraction example

- "The giant Snorlax will appear in the next week's Anipoke!" (Raishu no Anipoke wa Kabigon kyodaika desu.)
 - "Anipoke" and "Snorlax"
- "Pikachu is cute today too!" (Kyou mo Pikachu kawaii!)

"Pikachu" and "cute'



MMR method

Output

The MMR method is used because similar expressions are duplicated by simply extracting the phrase.

$$MMR := \underset{D_i \in R \setminus S}{\operatorname{arg\,max}} \left[\lambda \cdot Sim_1(D_i, Q) - (1 - \lambda) \underset{D_j \in S}{\operatorname{max}} Sim_2(D_i, D_j) \right]$$

Scene classification

Manually classify scenes into 6 categories.

- People: Scenes in which people appear
- Action: A scene in which a person takes action
- Emotion: A scene in which a person shows emotion
- Summary: Rough description of a scene without humans
- Scenery: A scene where the location and time of day can be observed Conversion: A change of scene from the main program to other images (e.g., a commercial)

5. Conclusion

Event scene search

The next deployment

- about 5% to 10% of the total error that was observed in the performance evaluation was due to the time
- MMR was able to suppress duplication of similar key phrases.
- Difference between correct answer data and tweet text.
- We used a Wikipedia-based trained model, but We assume We couldn't handle internet slang such as "Kita" well.

Use pre-learned BERT based on SNS

- Weighting key phrases
- Expansion of key phrase candidates by optimizing part of speech conditions

extracted.

- *This work was supported by JSPS KAKENHI Grant Numbers JP20K12027, JP18K11549.