

Time-Variable Analysis of Accommodation Reviews Based on Hierarchical Topic Model

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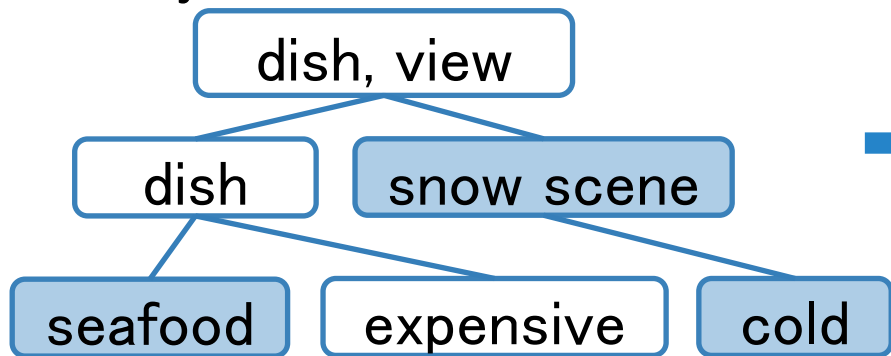
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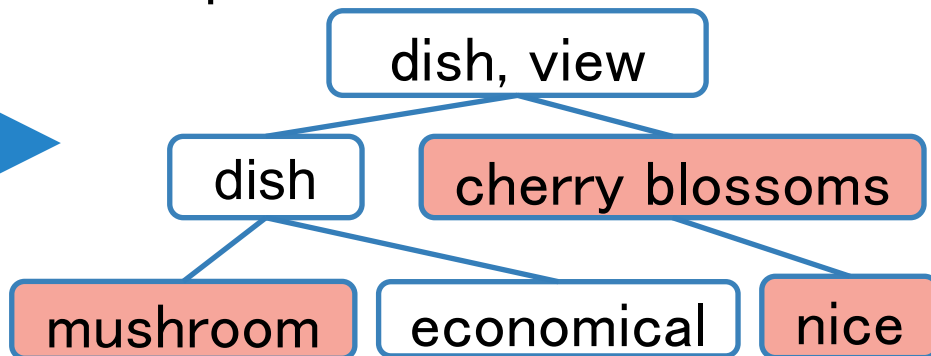
1 | Abstract

1. We propose a method to extract **seasonal features** from accommodation reviews.
2. We evaluated the change of the similarity among the features.

January



April



2 | Background

Popular accommodations have many reviews.
The required information cannot be found.

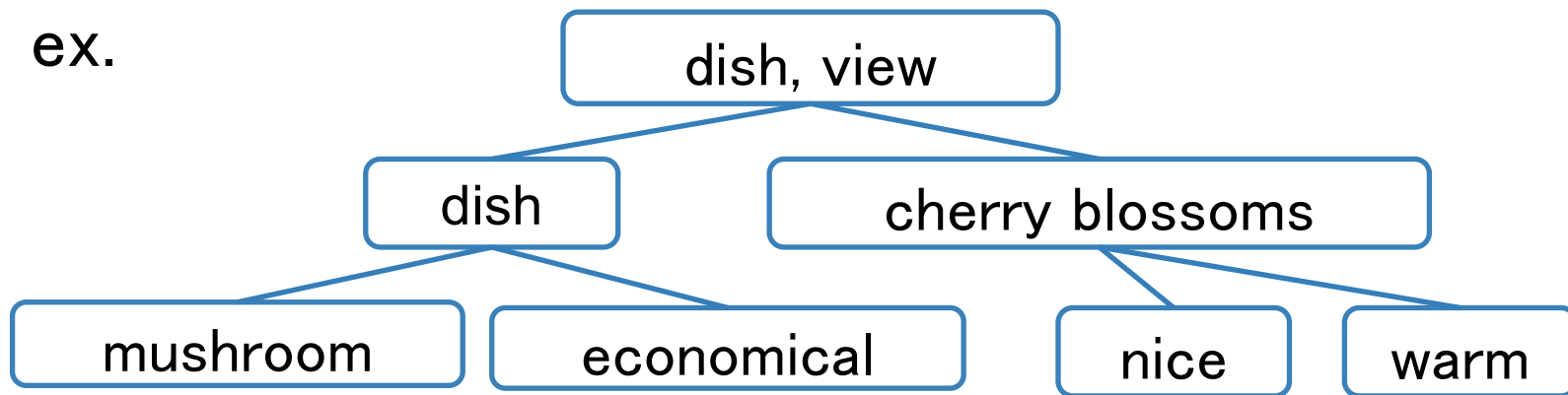
You can only see 8 or 7 reviews per page.

- It is not possible to discover seasonal changes and the characteristics of the season in which you plan to stay.

3 | Purpose of study

We extract the features of accommodation facilities in a **hierarchical structure**

ex.



1. Abstract

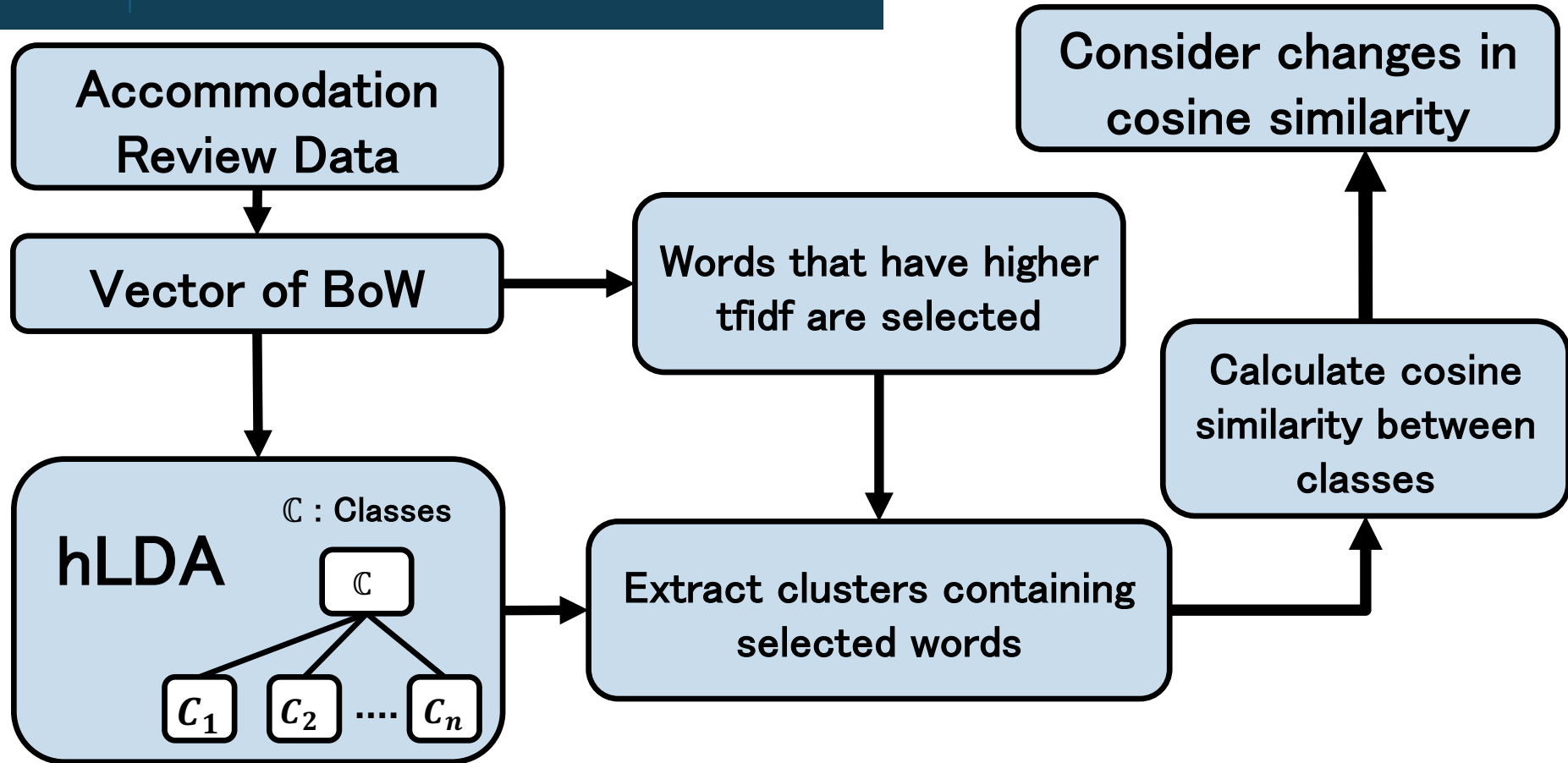
2. The proposed method

3. Result of analysis

4. Discussion

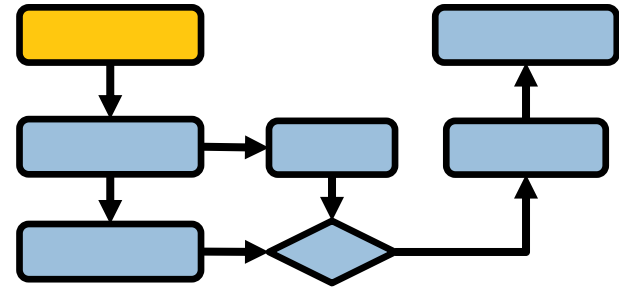
5. Conclusion

4 Framework of method



Rakuten Travel reviews

- During year of 1996 through 2016.
 - The top 10% of accommodations with the highest number of reviews.
-
- ◆ #1 Accommodations (3,118 reviews)
 - ◆ #2 Accommodations (2,770 reviews)



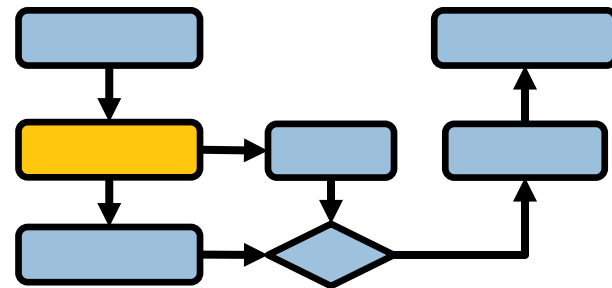
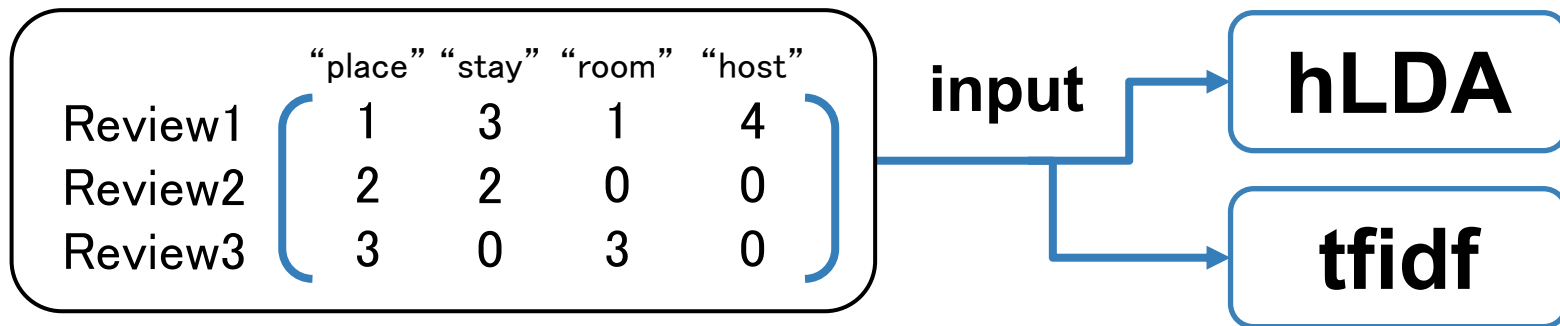
6

Text vectorization on BoW

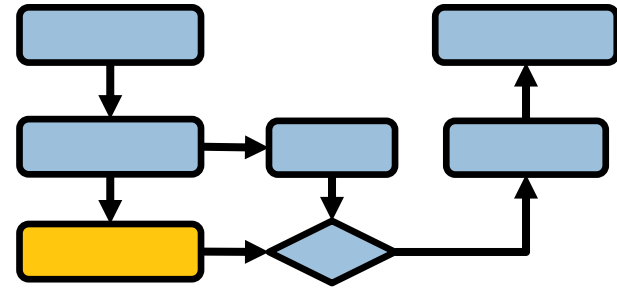
Splitting reviews by month



Vectorized based on the BoW



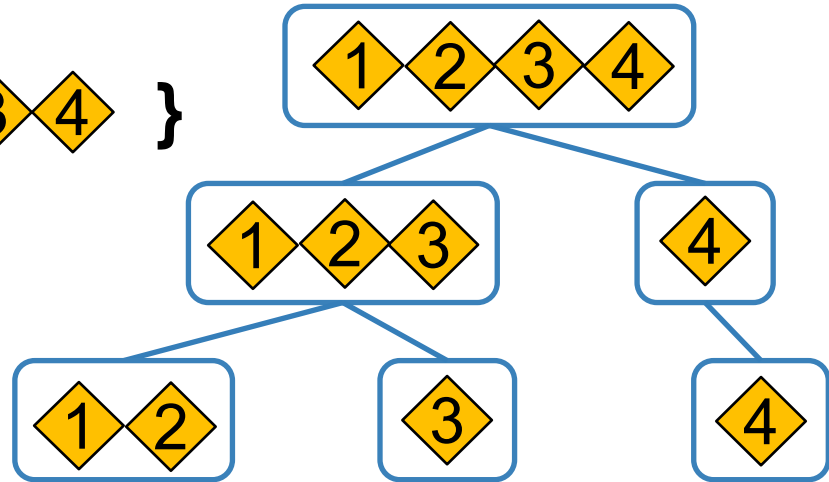
7 | hLDA(1/2)



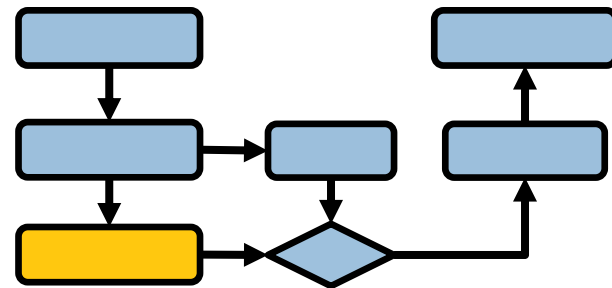
hLDA (hierarchical Latent Dirichlet Allocation)

- ✓ The hLDA analyzes the hierarchical relationship of topics.

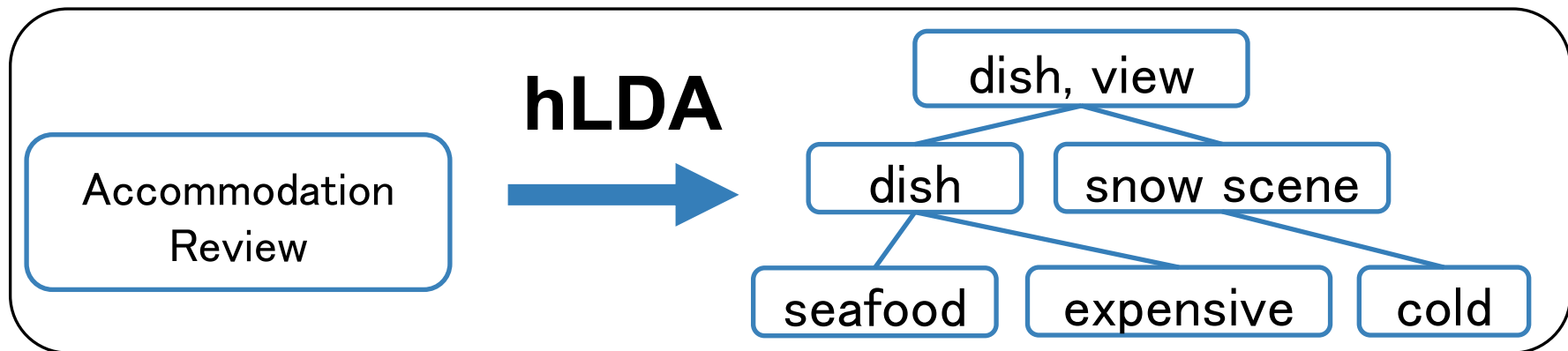
Topics = {  }



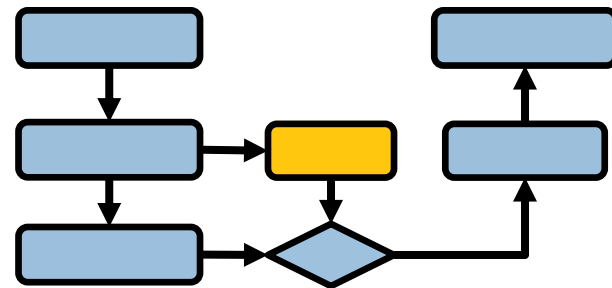
8 | hLDA (2/2)



We hypothesized that hLDA could be used to extract accommodation topics from accommodation reviews.

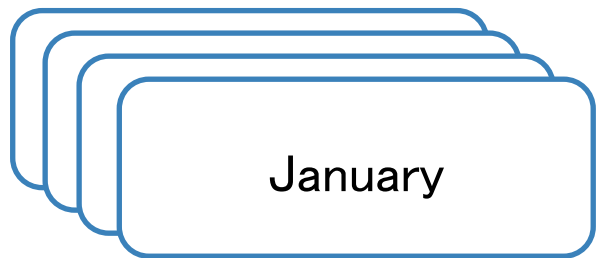


9 | TF-IDF



- ✓ TF-IDF can be used to compare the relative frequency of words in a document with the inverse ratio of words throughout.
- ✓ In this study, nouns with high values can be said to be **seasonal feature words**, because monthly documents are used.

Monthly Documents



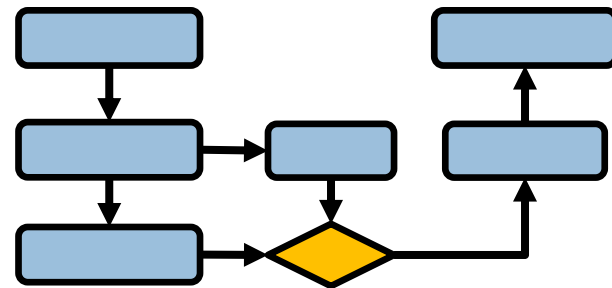
TF-IDF



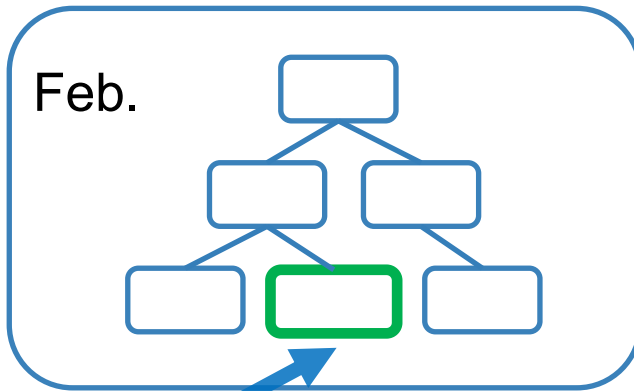
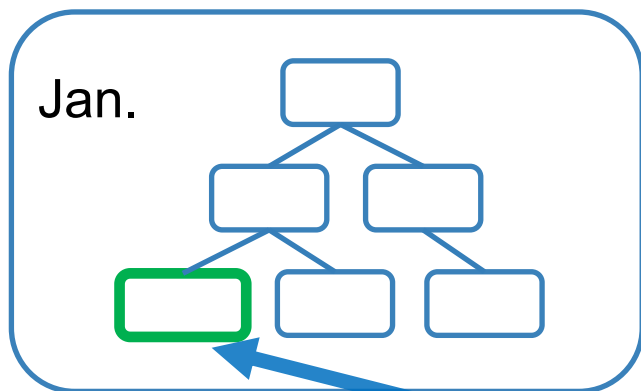
Seasonal feature words

The top 10% nouns
with TF-IDF values

10 Focusing on the clusters



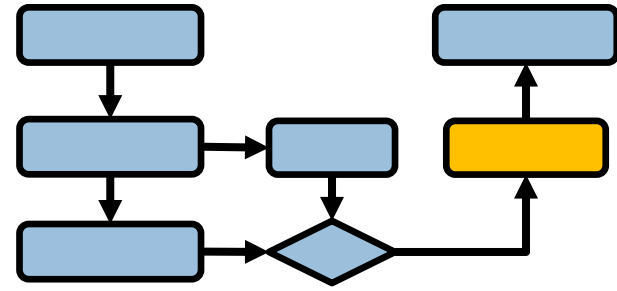
We extract clusters containing seasonal feature words



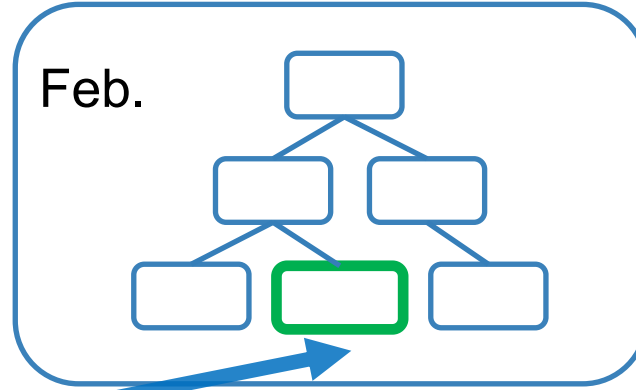
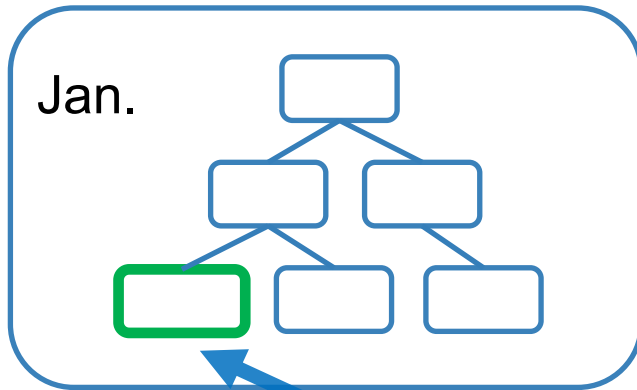
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Seasonal feature words extracted by the tf-idf values.

11 Cosine similarity(1/2)



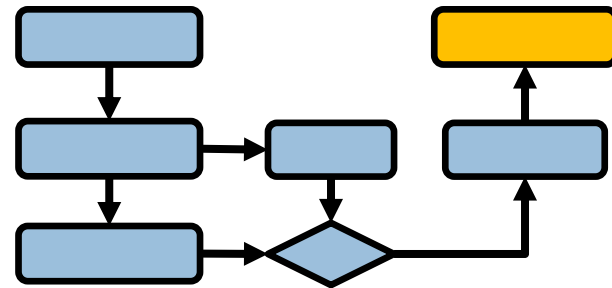
We evaluate the similarity between the clusters



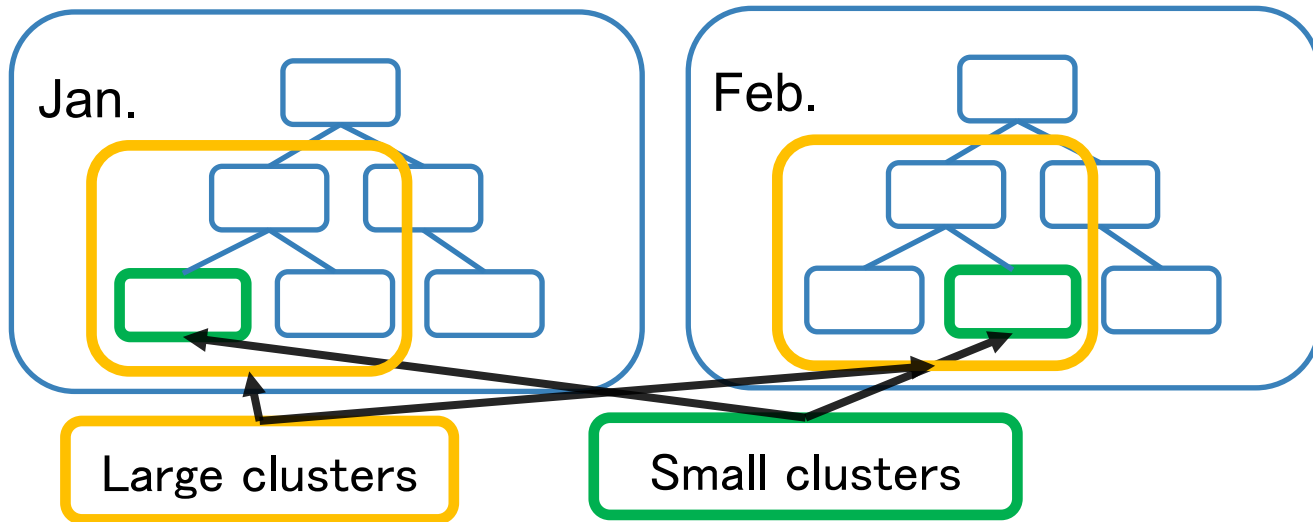
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High: The clusters have similar features.
Low: Seasonal features have changed.

12 Cosine similarity(2/2)



We evaluate the amount of change in similarity



✓ We extract seasonal changes from two types of clusters.

1. Abstract

2. The proposed method

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13 The result of hLDA(1/2)

The result of hLDA for accommodation #1 in January.

dish, room, crab, satisfaction, room,
waitress, bath, meal, hot spring

cleanliness, Kappou, hot pan, degree,
trip, hot spring, season, profit

snowy road, regulation, guest room,
effect, snow, body, condition

crab, dish,
field trip, crab,
sweet taste, fine,
Echizen, building

Kotatsu, Awaze,
hot spring, use,
overnight, city,
noisy

admiration, crew,
snow viewing, neat,
nature, once
consideration, first,

heating,
opinion,
summer
vacation,
overnight

14 The result of hLDA(2/2)

The result of hLDA for accommodation #2 in August.

room, hotel, use, stay, regret,
satisfaction, **view**, breakfast, great

rating, viking, **location**, **Maiko Station**,
neighborhood, checkout, price

reception, **view**, paying, impression,
beach, **sea**, bathing, ship, menu

towel, display,
atmosphere,
resort, air
conditioner,
sound, with family

convenience store,
room, crowded,
check-in, crowded,
annex, evaluation

this time, **sea**,
beach, feeling,
children, purpose

limited, whole
ward, fee,
customer, big
bath, type,
service, **view**

15 | The result of TF-IDF

Top 5 nouns with high TF-IDF values

✓ TRANSLATED INTO ENGLISH BY THE AUTHORS

	Accommodations #1	Accommodations #2
Jan.	new year, new year's day, new year's end, mochi pounding , superlative degree	new year, new year's, sweets, special, anniversary
May	Golden Week, red snapper , spring, love, holiday	room temperature, Golden Week, weekend, grade, beauty treatment
Aug.	Obon, Gassho, Rokusaburo Michiba, sweetfish , Noryo	pool , summer vacation, beach , sea bathing , barbecue
Dec.	christmas, meal, breakthrough, specialty, superlative degree	Luminarie , christmas, win- ter, hospitality, special

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16 Discussion (1/3)

Accommodations #1 focus on “Red snapper”

month	Jan.~ Mar.	Apr.	May	June	July	Aug.	Sept.	Oct.	Nov.~ Dec.
Large	0.0	0.029	1.0	0.046	0.018	0.018	0.021	0.028	0.0
Small	0.0	0.099	1.0	0.199	0.099	0.099	0.099	0.105	0.0

- ✓ People can enjoy “Red snapper” from April to October
- ✓ Similar features can be seen from April to June

Nouns appeared with “Red snapper” were cuisines: KAISEKI (Traditional multi-course meal), nice smell, CHAZUKE (rice in Japanese broth).

17 Discussion (2/3)

Accommodations #2 focus on “pool”

month	Jan.~ May	June	July	Aug.	Sept.	Oct. ~ Dec.
Large	0.0	0.110	0.042	1.0	0.018	0.0
Small	0.0	0.099	0.099	1.0	0.099	0.0

- ✓ Reviews changed drastically due to low similarity in Jul. and Sep.
- ✓ Many families instead of single travelers stay at the accommodation in August.

Aug. cluster includes: Together, pool, Children, adults, travel ...

18 Discussion (3/3)

Accommodations #1

- ✓ “Red snapper” is popular from April to June
- ✓ Guests have a strong interest in food.

Accommodations #2

- ✓ Family trips are popular in August
- ✓ Guests can use swimming pools from June to September.

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19 Conclusion

1. We propose a method using hLDA to extract seasonal features from accommodation reviews.
2. We verified the effectiveness of the proposed method using TF-IDF and cosine similarity.
3. We found a seasonal difference in each cluster.
4. As the future work , we will differences in features in the region.