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Exploring Relevant Parts between Legal Documents using Substructure Matching

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Ordinances and Rules in Local Governments (OR documents for short)

- Regulations for social lives
- In Japan, about a million in 1,788 local governments
- In wide range



harmful contents for kids



drunk-driving



environmental pollution

Drafting OR Documents

- Officers draft according to social situations.
 - 1. Search relevant existing OR documents,
 - 2. imitate them for the first draft, and
 - 3. modify the content for the situations.
- Questions
 - How officers search *relevant* OR documents?
 - Heuristics (e.g., those of "similar" governments, popular ones, etc.)
 - How they choose which *parts* of the documents for imitation?
 - On manual

What we can do?

- OR document search \approx document similarity search
 - Vectorization (BoW, TF-IDF, Doc2vec, etc.)
 - Topic modeling (LSI, LDA, etc.)
 - Similarity computation (cosine similarity, Tversky index, etc.)
- Parts determination for imitation
 - Idea: different parts (e.g., sentences) are missing in the drafting OR document.
 - \rightarrow diff can be a choice.
 - However, there can be too many different sentences.
 - Even one character difference.
 - Moreover, irrelevant provisions would be included.

Various Granularity of Provisions

- Dependent on individual local governments.
 - e.g., Protection of young persons ordinances



"Same" provision = Same content? – No

- Different governments have different contents.
 - Different structure and ordering, missing contents, etc.

Paragraphs about council in	Landscape ordinances
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Ami			Shichigashuku		
#A	#P	Article Title (Content desc.)	#A	#P	Article Title (Content desc.)
23	1	Establishment	11	1	Council (Establishment)
24	1	Deliberation matters	11	2	Council (Deliberation matters)
25	1	Counsel	11	3	Council (Counsel)
26	1	Organization ($\#$ Committee)	12	1	Organization ($\#$ Committee)
26	2	Organization (Conditions)	12	3	Organization (Conditions)
26	3	Organization (Temporary com.)			
27	1	Term of service (Basic)	19	2	Organization (Torm of sorvice)
27	2	Term of service (Substitute)			Organization (Term of Service)
27	3	Term of service (Temporary com.)			
28	1	Chairperson (Election)			
28	2	Chairperson (Chairperson)			
28	3	Chairperson (Vice chairperson)			
29	1	Convention (Summons)			
29	2	Convention (Resolution)			
30	1	Section			

Objective

- Intuition
 - Missing contents in "same" provisions are helpful to determine which parts to imitate.

Ami			Shichigashuku		
#A	₩P	Article Title (Content desc.)	#A	#P	Article Title (Content desc.)
23	1	Establishment	11	1	Council (Establishment)
24	1	Deliberation matters	11	2	Council (Deliberation matters)
25	1	Counsel	11	3	Council (Counsel)
26	1	Organization (#Committee)	12	1	Organization ($\#$ Committee)
26	2	Organization (Conditions)	12	3	Organization (Conditions)
26	3	Organization (Temporary com.)			

Questions

- Which parts are corresponding with "same" provisions?
- How to determine missing contents?

Idea: Substructure matching by regarding OR documents as tree-structured data

- OR documents as tree-structured data
 - Subtree is regarded as provision





Idea: Substructure Matching

• Same provision: subtree pair having same contents



• Relaxations: similarity matching



Proposed measurement: Matching Ratio

• Intuition:

the more matched contents, the more similar provisions

$$R(T_i, T_j, M_i, M_j) = \sqrt{\frac{|M_i|}{|S_i|} \cdot \frac{|M_j|}{|S_j|}}$$

 M_i : set of matched contents in T_i S_i : set of leaf nodes in T_i

• Example on the right

•
$$R(T_1, T_2, M_1, M_2) = \sqrt{\frac{5}{8} \cdot \frac{5}{8}} = 0.63$$



Same Matching Ratio, but ...

• Matching ratio cannot distinguish the following.



Proposed measurement: Provision commonality

• Intuition:

the larger number of consecutive non-matched leaf nodes, the more different contents are included.

$$C(T_i, T_j, M_i, M_j) = \sqrt{\frac{1}{L(T_i, M_i) + 1} \cdot \frac{1}{L(T_j, M_j) + 2}}$$

 $L(T_i, M_i)$: the longest consecutive non-matched leaf nodes in T_i



Provision commonalities



Search Algorithm (see paper for detail)

- Given:
 - Two OR documents: T_i , T_j
 - Contents matching: M
- Find:
 - Vertex pairs $P \subseteq T_i . V \times T_j . V$ s.t. linear combination of $R(\cdot)$ and $C(\cdot) >$ threshold
 - P is maximal to avoid redundancy
 - Vertex pair $(v, u) \notin P$ s.t. ascending vertex pairs of v, u are in P
- Idea
 - Bottom up manner
 - Eliminate candidate pairs if they cannot be in *P*.

Evaluation

- Settings
 - Survey by Ito [6]
 - Inclusions of specific classes of provisions (e.g., council and prohibitions) in articles of the landscape ordinances.
 - Examination
 - Check whether discovered pairs belong same class.
 - Baseline approach is same as the proposed method except focusing only on articles.
 - Evaluation metrics: Precision, Recall, F1-score
- Results

(a) Optimistic.

Method	Precision	Recall	F1-measure
Baseline	0.34	0.81	0.48
Proposed	0.51	0.79	0.62

including "others" class

(b) Skeptical.

Method	Precision	Recall	F1-measure
Baseline	0.39	0.83	0.53
Proposed	0.37	0.83	0.51

excluding "others" class

Use Case

Paragraphs about council in Landscape ordinances

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Conclusion

- Motivation: support for OR document drafters
 - Searching relevant OR documents
 - Discovering parts for imitation
- Approach: relevancy metrics
 - Matching ratio: content matching
 - Provision commonality: structural matching
- Result
 - 0.62 F1 score for finding matches
 - Useful use case