

A Study on Descriptive Patterns Based on Similarity Classes of Individual Constants

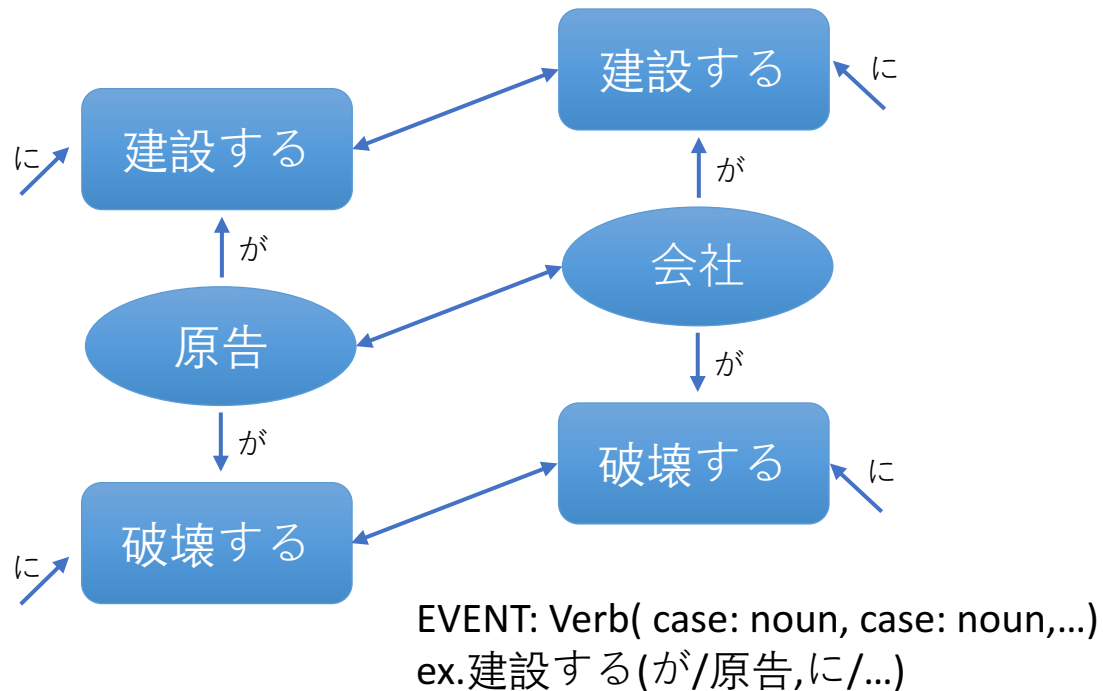
個体の類似クラスに基づく記述的類似性

Knowledge Base Lab.

M2

Ruipeng Wang

Descriptive similarity

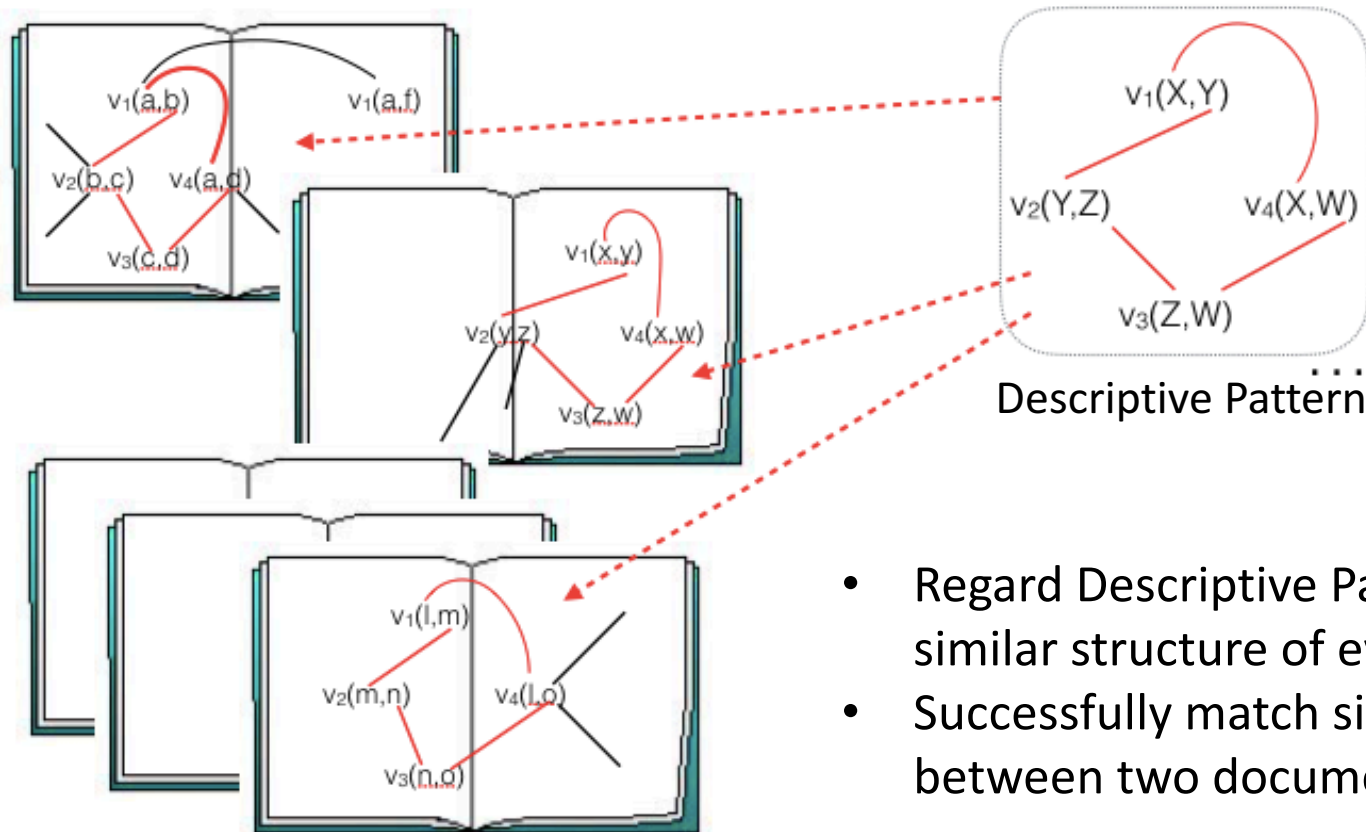


One to one correspondence :

Nouns come with same verb-cases from different sentences

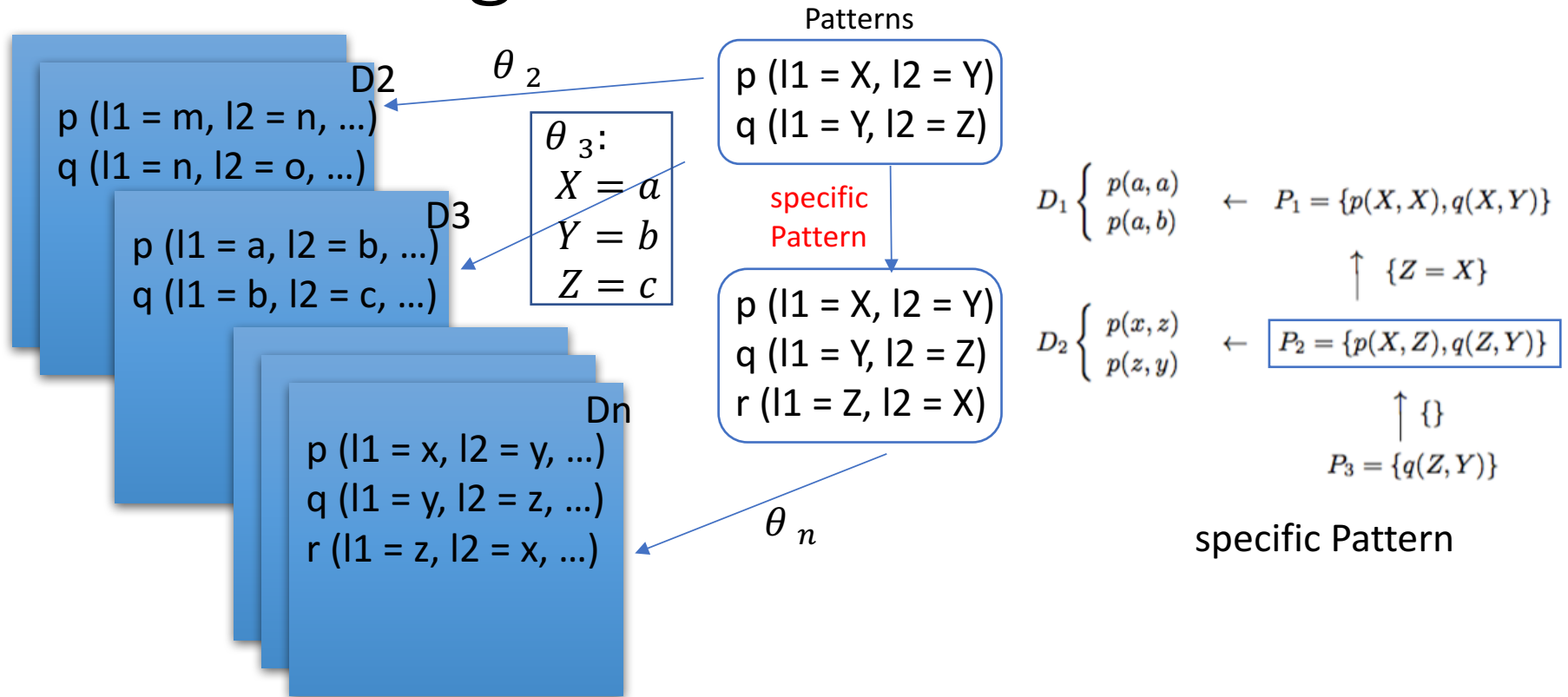
Previous research

X. Zhang, Feb. 2016, Master Thesis



- Regard Descriptive Pattern as a similar structure of event graphs.
- Successfully match similar structure between two documents

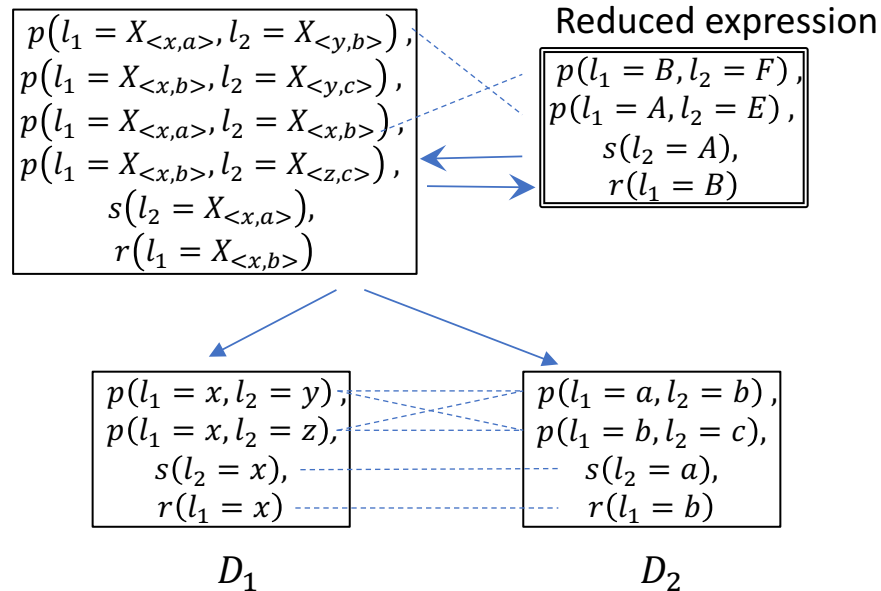
Research target



- Regard group of event as patterns, and extract most specific frequency pattern as descriptive pattern.
- Consider multi-document situation, extract patterns supported by multi-document. (a kind of Data Mining instead of matching.)
- Phase 1: extract similarity class of Individual Constants. (Input of next phase)
- Phase 2: Re-construct descriptive patterns. (Depend on the result of Phase 1)

Least General Generalization

LGG process



- Time complexity will be very large will computing LGG.
- We propose to use neither pairs nor tuples: Similarity classes of individuals over domains: frequent closures (intent of formal concepts)

Maximal closures

role set with maximal roles is complete corresponding with DP

Descriptive Pattern, P:
 $\{ V_0(l_1 = X, l_2 = Y, \dots), V_1(l_1 = Y, l_2 = X, \dots), \dots \}$

verb-role pairs played by the variable X

$$role_P(X) = \begin{cases} \text{maximal:} & \{v_0(l_1), v_1(l_2)\} \\ \text{non-maximal:} & \{v_0(l_1)\}, \{v_1(l_2)\} \end{cases}$$

$a \dots$	$x \quad y$	$\cdot m$
$const(D_1)$	$const(D_2)$	\dots

$D_1 \qquad D_2 \qquad D_3$

Inner Predicate composition:
 recover the co-occurrence information of variables.

$v_0(l_1 = X), v_2(l_2 = Y)$
 $v_1(l_1 = Y), v_2(l_2 = X)$

Pattern whose event has just one role description

- descriptive pattern can be re-constructed from maximal closure


Inner Predicate composition


Any variable in DP has a closure: $A = \{v_0(l_1), v_1(l_1)\}$

Conversely we make primitive patterns from them, and
compose the primitive patterns to get predicates with more arguments

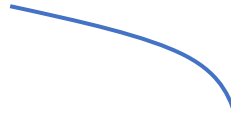
Primitive pattern for closures

$$pp(A) = v_0(l_1 = A), v_1(l_1 = A)$$


$$v_1(l_2 = B), v_2(l_1 = B)$$


$$v_2(l_2 = C)$$

$$v_3(l_1 = D)$$


$$v_4(l_1 = E), v_3(l_2 = E)$$

Inner predicate composition

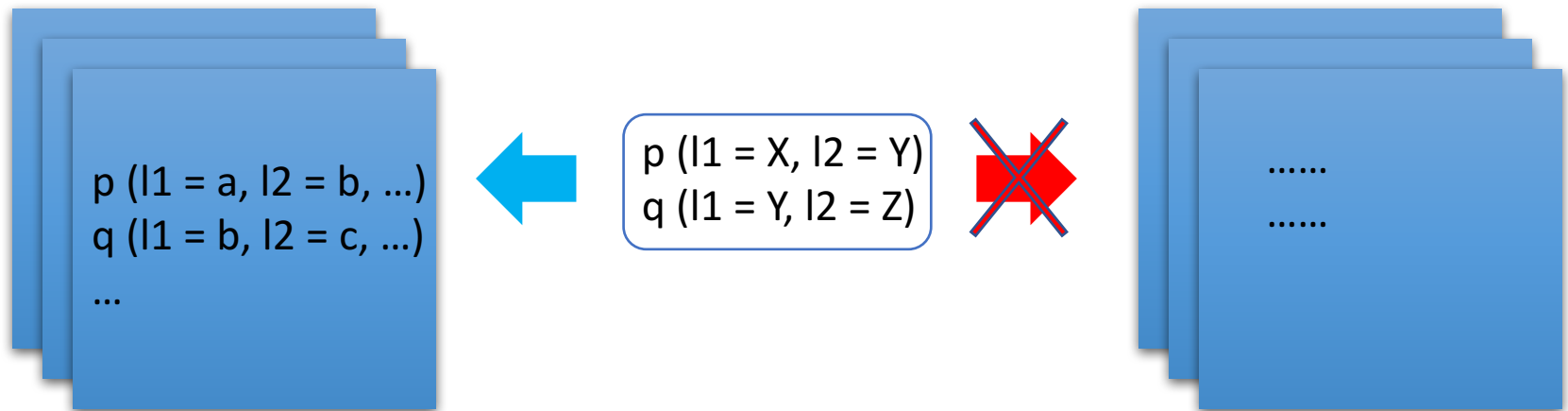
$$v_0(l_1 = A), v_1(l_1 = A, l_2 = B), \\ v_2(l_1 = B, l_2 = C)$$

$$v_4(l_1 = E), v_3(l_2 = E, l_1 = D)$$

More specific pattern towards DP

- Beam Search Algorithm to extract some main descriptive patterns.

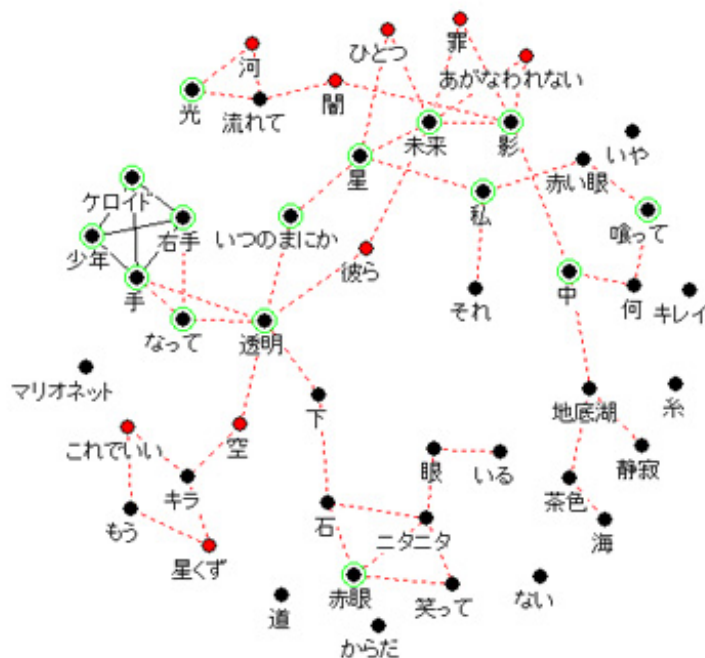
Requirement: Minimal Support



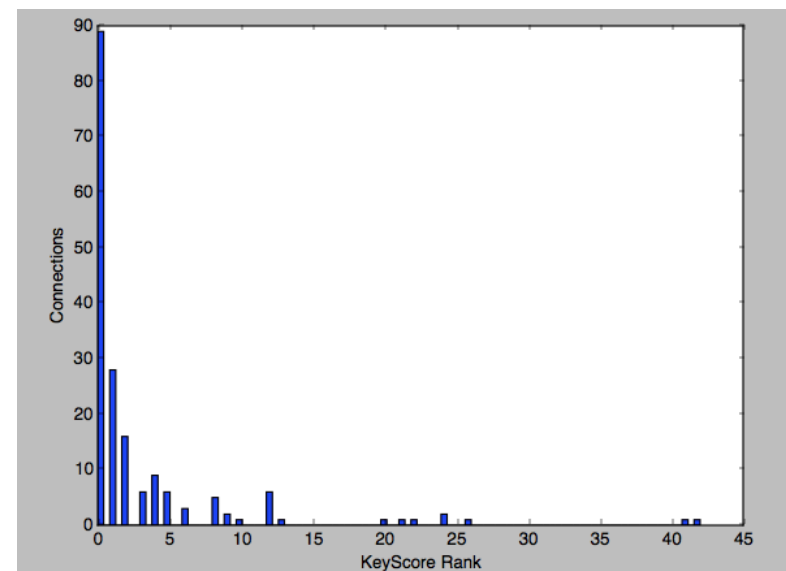
- a pattern is supported by multi-document.
- We set τ donates the percentage of document that support a pattern.
- ex. For a Pattern P , $\tau = 1$ means all documents support P .
- Our pattern is required to be supported by multi-document.

Requirement: KeyGraph

- Closures to be candidate include at least one High KeyScore noun.
- KeyScore reflects the importance of a noun in a document.
- Feasibility: connections between different events mainly generated by high KeyScore nouns.



black: high frequency words (hf)
red : not hf but close contact to hf
green : high KeyScore words

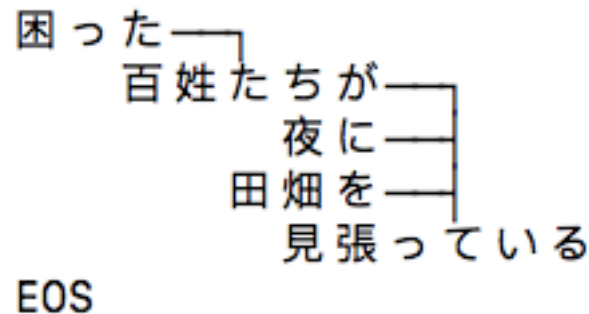


KeyScore Rank words over 45th are omitted

Data

- We can use KNP tool to extract events with verb-case information.

- 困った百姓たちが夜に田畑を見張っている



- 見張る(百姓/ガ, 夜/ニ, 田畑/ヲ)
- We use 100 precedents and extract 84,550 events.
- Average length of the precedents is 45,000 Japanese words.

Experiment: Stories

data: two short Japanese stories, $t = 1$.

part of closures extracted

- * [荒らす/ヲ] [町001, 八百屋001, 田畑002]
- * [持つ/ヲ] [光001, 火縄銃002]
- * [困る/カ] [若者001, 村人002]
- * [現れる/カ] [化け物001, 鹿002, 牛001, 老人001]

original story 1

...ところがある夜から、金色の二つの光を持った化け物が現れ、町の八百屋を荒らして回るようになった。困った若者たちが、夜の八百屋を見張っていると...

original story 2

...夜な夜な2頭のつがいの大鹿が現れ、田畑を荒らしまわるので、村人は大層困っていた...

あやしい牛	あばれ鹿
町	田畑
若者	村人
光	火縄銃
牛	鹿

Experiment: Precedents

data: three precedents include two similar ones and a very different one, $t = 2/3$

words(noun)	Event(s)	KeyGraph(s)	MFC(s)
76083	0.15	10.31	8.27

105 maximal closures founded by these three document, and **98** closures are supported by the similar ones. Over **93%**.

違反 ガ/被告068, 原告092 ニ/義務068 条092
規定 ガ/義務068 義務092 ニ/義務068 条092
履行 ガ/被告092 被告068 ヲ/義務068 義務092
主張 ガ/被告092 被告068 ヲ/原告068 士092
受ける ガ/原告068 士092 カラ/国068 原告092
開設 ニ/国068 原告092 ガ/県092 被告068
持つ ガ/県092 被告068 ヲ/家族068 責任092

part of descriptive pattern re-constructed by extracted maximal closures

Summary and Future Work

- Give the exact definition of Descriptive Pattern.
- Validate the feasibility of KeyGraph.
- Extract similarity classes from multi-document.
- Realize the re-construct descriptive pattern from maximal closures.
- Adjust the output of KNP, fix the results.