

Cross-Language IR at University of Tsukuba

Automatic Transliteration for Japanese, English, and Korean

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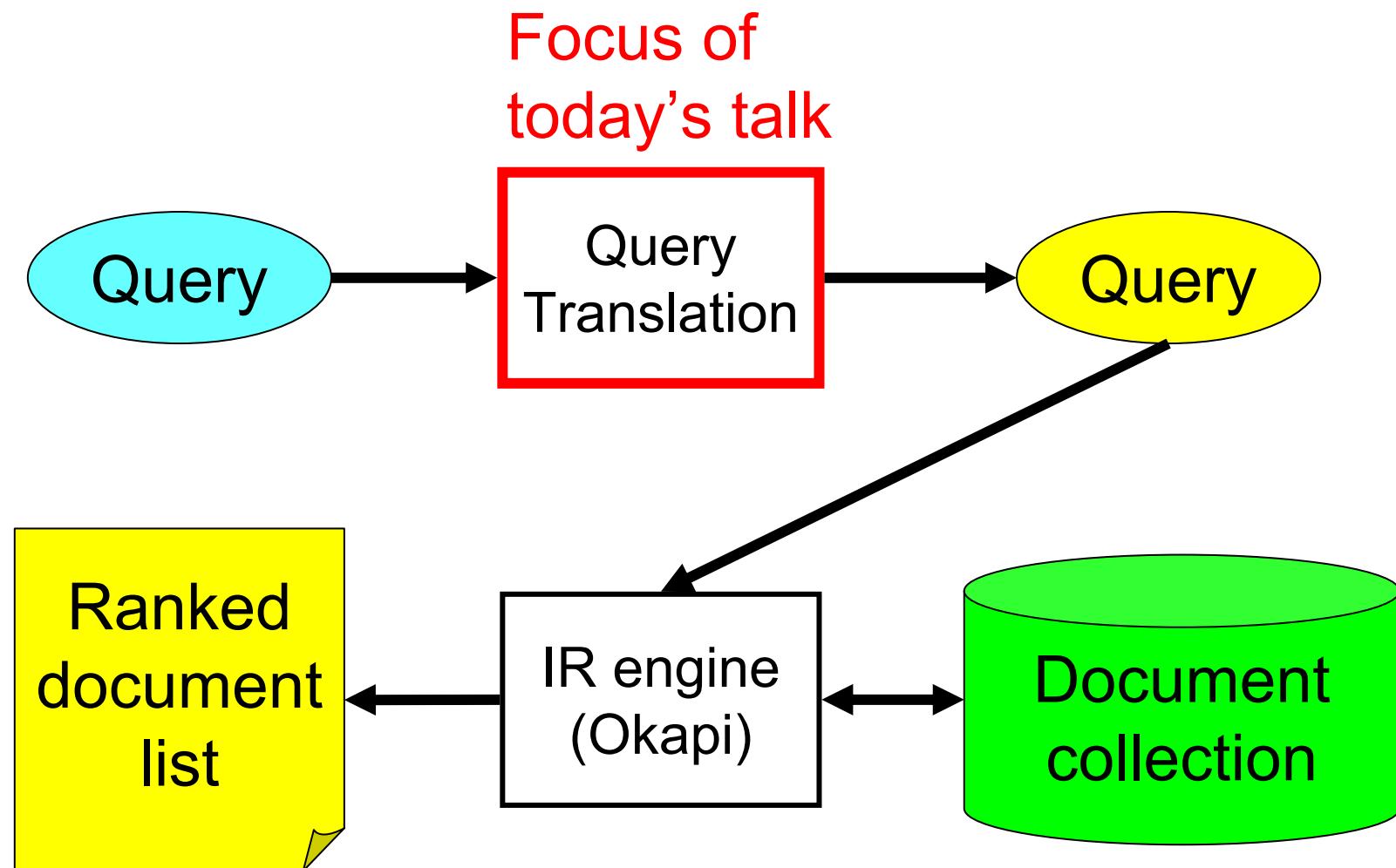
Motivation

- We developed an automatic **transliteration** method for Japanese and English CLIR
 - effective in translating foreign words spelled out by phonetic alphabet (e.g., Katakana)
 - evaluation since NTCIR-1
 - the method has been used in commercial cross-language patent IR service
- In NTCIR-4 CLIR, we applied our method to Korean and **realized JEK transliteration in a single framework**

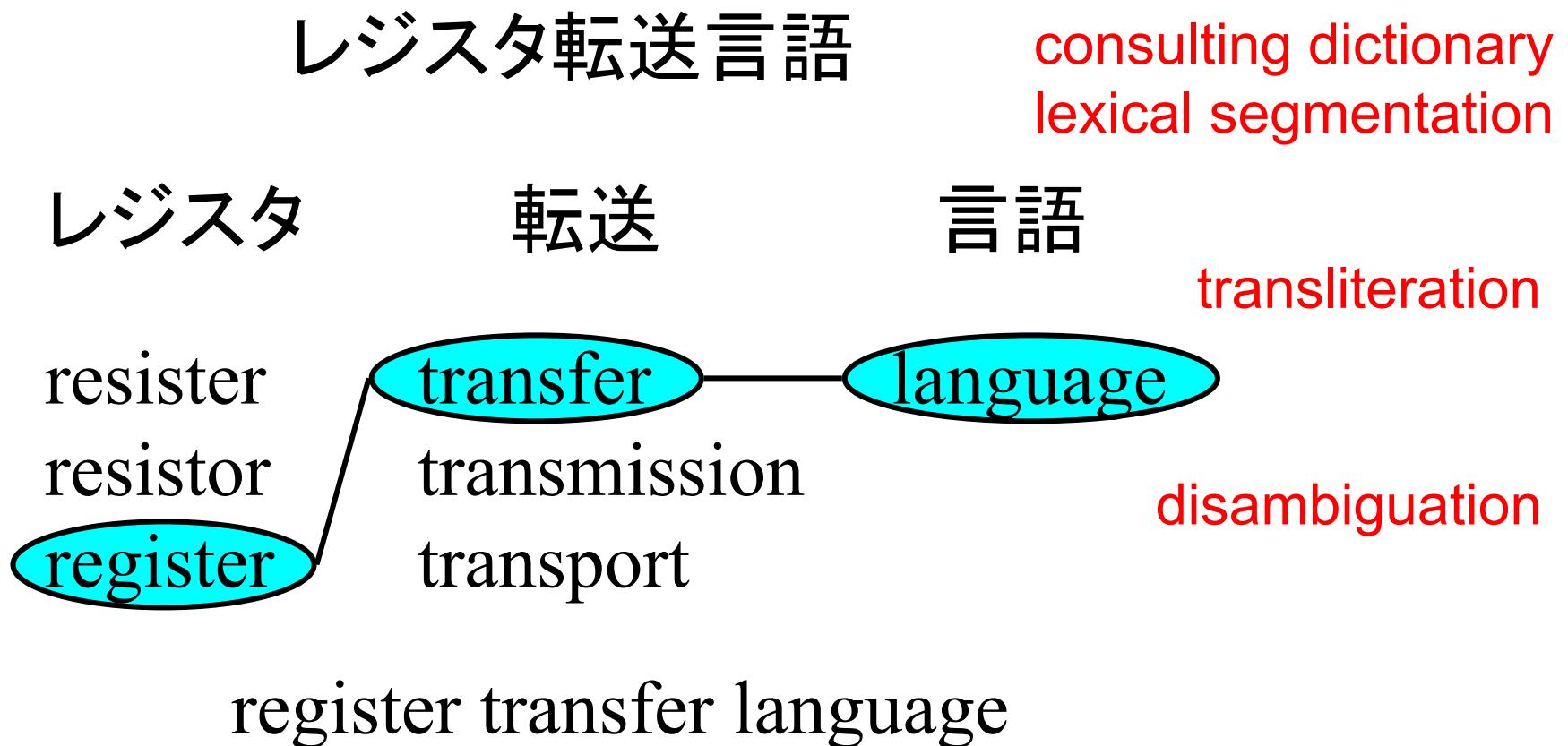
Basis of transliteration

- spelling out foreign words (loanwords) by phonetic alphabet
 - technical terms and proper names
 - often out-of-dictionary words
- examples
 - dioxin → ダイオキシン, 다이옥신
 - Yugoslavia → ユーゴスラビア, 유고슬라비아
- back-transliteration
 - process to identify the source English word

Overview of our CLIR system



Example of J-E Query Translation



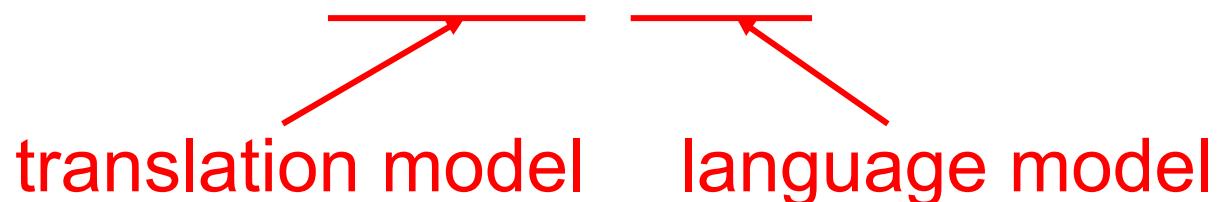
Query Translation (cont.)

- compound query term S and a translation candidate T

$$S = s_1, s_2, \dots, s_N \quad \text{si and ti are base words}$$

$$T = t_1, t_2, \dots, t_M$$

- compute $P(T|S) = P(S|T) \cdot P(T)$



- select the candidate with max $P(T|S)$

Translation model

- $P(S|T) = \prod P(s_i | t_i)$
 s_i and t_i are base words comprising S and T
- heuristics and EM algorithm to correspond dictionary entries on a word-by-word basis

情報検索システム	Information retrieval system
検索モデル	retrieval model
情報抽出システム	Information extraction system
特許情報処理	patent information processing

- estimate $P(s_i | t_i)$

Language model

- word-based trigram model
- 100K vocabulary in a target document collection
- Palmkit was used
 - compatible with CMU-LM toolkit

Transliteration method

- out-of-dictionary word S and a transliteration candidate T

$$S = s_1, s_2, \dots, s_N$$

$$T = t_1, t_2, \dots, t_M$$

si and ti are letters (substrings of words)

- compute $P(T|S) = P(S|T) \cdot P(T)$

$$\overbrace{P(S|T) \cdot P(T)}^{\begin{array}{l} \text{transliteration model} \\ \text{language model} \\ (\text{word unigram}) \end{array}}$$

- select the candidate with max $P(T|S)$

Transliteration dictionary

- dictionary for transliteration includes correspondence b/w source and target words on a phonogram-by-phonogram basis
- we use Roman representation as a pivot

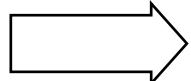
Producing J/E dictionary

1. extract Japanese **Katakana** words and English translations from J-E dictionary
2. romanize Katakana words
3. correspond romanized Katakana and English words on a letter-by-letter basis
4. find the best correspondence

Example matrix

テキスト(te-ki-su-to) text

	テ	キ	ス	ト	\$
t	3	1	2	3	0
e	0	0	0	0	0
x	1	2	1	1	0
t	3	1	2	3	0
\$	0	0	0	0	3



テ te
キス x
ト t

By performing the same process for all Katakana entries, we produce transliteration dictionary

Extension to other languages

- our transliteration method can be applied to any language **if represented by Roman characters**
- no existing method has been used and evaluated in CLIR for more than two languages
 - our experiment was the first effort to explore this issue

Problems in Korean

- romanization of Korean words is more difficult than that of Katakana words
 - # of Hangul characters is approx. 11,000
 - one-to-one mapping b/w Hangul and Roman characters is not easy
- both conventional Korean words and foreign words are written by Hangul characters
 - detection of foreign words in Korean dictionary is crucial

Romanizing Korean words

- Hangul character consists of three types of consonants
 - first consonant (19)
 - vowel (21)
 - last consonant (27 + 1)
- # of possible combinations is 11,172
(# of common characters is approx. 2,000)
- We used Unicode, in which characters are coded according to consonants

last consonant is optional



Fragment of Unicode table

- first consonant changes every 21 lines
 - vowel changes every line and repeats every 21 lines
 - last consonant changes every column

- Hangul characters can be identified by pronunciation
 - only map b/w consonants and Roman characters is needed

Fragment of Unicode table

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ㄱ ㄲ ㅋ ㄴ ㄷ ㄸ ㄹ ㅁ ㅂ ㅃ ㅅ ㅆ ㅇ ㅈ ㅊ ㅋ ㅌ ㅍ ㅎ	
가	44032: 가각깎간간간간간갈깎갈깎갈깎갈깎갈깎감갑깎갓갓강갓갓갓같각같같같 44059
개	44060: 개객걔걔걔걔걔걔깰깰깰깰깰깰깰깰깰깰깰깰깰깰깰깰깰깰깰 44087
까	44620: 까깍깎깎깎깎깎깎깎깎깎깎깎깎깎깎깎깎깎깎깎깎깎깎 44647
깨	44648: 깨깱깱깱깱깱깱깱깱깱깱깱깱깱깱깱깱 44675
나	45208: 나낙낙난난난난난난난난난난난난난난난난난난난난난 45235
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Detecting foreign words in Korean

- compute the phonetic similarity b/w romanized Hangul words and their translations (either English or Japanese)
- discard translation pairs whose similarity is below a threshold
 - conventional Korean words are discarded
- foreign word entries remained

Experiments (J/E)

<TITLE>, mean average precision (rigid)

Languages	#Entries	w/o transliteration	w/ transliteration
J-E	1M	0.2174	< 0.2182
E-J	1M	0.1250	= 0.1250
J-E (EDICT)	108K	0.1147	< 0.1383
E-J (EDICT)	108K	0.0612	< 0.0857

transliteration was effective for small dictionaries

Experiments (Korean)

<TITLE>, mean average precision (rigid)

Languages	w/o transliteration	w/ transliteration
J-K	0.2177	< 0.2457
K-J	0.1486	< 0.1746
E-K	0.2026	< 0.2153
K-E	0.1017	< 0.1231

transliteration was also effective for Korean

Conclusion

- realized transliteration for Japanese, English, and Korean in a single framework
- evaluated its effectiveness in NTCIR-4 CLIR task