**NTCIR-4 Chinese, English, Korean Cross Language Retrieval Experiments using PIRCS**

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**Abstract**

In NTCIR-4 we participated in Korean, Chinese and English monolingual retrieval, Chinese-English, English-Korean bilingual, and Chinese-Korean cross language (using English as pivot language) retrieval tasks based on our PIRCS retrieval system. The query translation approach was employed for CLIR. We combined two MT translations for Chinese-English, and two for English-Korean. For the latter, a web-based entity-oriented translation procedure was also used to translate un-translated terms. Concatenation of MT output was found to lead to better CLIR effectiveness than single MT, while entity translation brings further improvements of about 15%. The direct bilingual CLIR perform between 71% and 88% of monolingual, and appear to be best among submissions. For retrieval from Korean collection, bigram performs better than word indexing, and combination of the two provides better results, in most cases. Chinese-Korean retrieval runs via English as pivot language provide results with mean average precision between 56% and 66% of our Korean monolingual runs. All submissions are automatic runs without manual intervention.

**Keywords**: monolingual Korean, Chinese retrieval; CLIR; MLIR; bigram indexing; short-word indexing.

1 Introduction

Participants in CLIR tasks need to experiment with more than two languages in NTCIR-4. We took this opportunity to add Korean (K) as the third language to our PIRCS retrieval system’s usual English (E) and Chinese (C) capability. Usage of these three languages is diagrammed in Fig.1 above to show the tasks that we have done and submitted. Our convention is to denote the query language via the notation $Q^{ABC}$: meaning that the final language is C and it has been derived through query translation from source language A via a pivot language B. Direct query translation is denoted as $Q^{AB}$, for example. The last superscript language character always indicates what collection language this query would operate on.

There were a total of fourteen runs. These include retrievals with Chinese target collections named as:  
pircs-C-C-T-01  pircs-C-C-D-02;  
pircs-C-E-T-01  pircs-C-E-D-02;  
pircs-C-K-T-01  pircs-C-K-D-02  
pircs-C-K-DN-03  pircs-E-K-T-01  pircs-E-K-D-02  
pircs-C-K-DN-03  pircs-C-K-DN-03

The pircs-C-K-xxx CLIR experiments employ the $Q^{CEK}$ queries with transitive translations. All retrievals include using the title or the description sections of the topics provided. In addition, for K-K and C-K experiments, runs using the description plus narrative sections are also submitted. We continue to use our PIRCS retrieval system [1]. This has been modified to support Korean processing. All our retrievals are automatic with PRF (pseudo-relevance feedback) as a default.

2 Translation Resources

The most important tools for cross language tasks are translation resources. We continue to employ the efficient query translation approach. Resources are needed to translate from Chinese to English, English to Korean and Chinese to Korean. The latter however seems not available easily (in the U.S.). Both Chinese
to English and English to Korean translation are new to us. These are considered major languages. We decide to use commercially available MT software for this purpose. We assume that they will provide reasonable translation for general English, but may not be sufficient for entity or terminology words. We augment the result with an entity/terminology-oriented web-based translation methodology that was being developing. One of our goals in NTCIR-4 is to test whether combining multiple MT outputs for query translation works better than single MT.

2a Chinese-English MT Software

For Chinese-English translation, Systran [2] and Loto [3] software packages were used. Systran has a long history of C-E translation. Loto is a product newly marketed in America; it evolved from the Huajian English-Chinese MT software in China. A license to Loto allows one to have a stand-alone MT package on a PC, as well as web access to their company’s central translation software. The latter is advertised to get updated regularly to provide better translation than the static, stand-alone version; but it is restricted to one single installation computer only. We have used the online translation facility for these experiments.

Our hypothesis is that combination of MT translation can bring more robust results. Given a query, two separate translations are performed and the results are concatenated together. If a source word/phrase leads to the same (duplicate) target translations, they may be regarded as ‘confirmed’ correct and are automatically weighted heavier. When translations differ, there is also possibility that they provide different wordings for the same source concept and therefore may hedge against insufficient coverage. In the case of English-Korean, one MT may provide semantic translation while the other may be more literal. The trade off is that when both were wrong, we end up with twice as much noise.

The following shows example output of typical Chinese-English translations of the description section of a topic for readers to judge their quality. Included at the end are six additional terms that are obtained from pre-translation expansion processing (see Section 4c).

$qry\#55$ Original Chinese: 亞洲各國對北韓發射大浦洞（Daepodong）飛彈反應，試射 彈道 和南 射程 防衛廳 飛越

$qry\#55$ English Translation via Systran: The Asian various countries launches the big water’s edge hole to North Korea (Daepodong) the guided missile response.

$qry\#55$ English Translation via Loto: Asian various countries launch the big Pu hole to Korea (Daepodong) The stray bullet reacts.

In general, there are both translation successes and failures. Except for entity names, the output appears acceptable for CLIR, both from the view of segmentation and translation.

2b English-Korean MT Software

In the U.S., resources for Korean language are not as common as other major languages. For English-Korean, we employed the English to CJK capability of Systran. Another package called English Guide (EnGuide) [4] from LniSoft was also acquired from Korea. The latter has user interface in Korean only, and is therefore not suitable for users who do not understand Korean. It also has difficulty handling sentences having words with capitalized first letter in the middle of a sentence (which is the case with the title section of our queries in English). We overcome this problem by producing two versions of the title section: one with the title all in lower case, and the other with the case information retained but put each word into a separate line.

Some examples of English-Korean translations are shown below. Included are un-translated English words that are picked up by our entity-oriented translation procedure (Section 2c).

$qry\#2$ Original English: Find out who joined the Jonnie Walker Charity Golf Tournament in Taiwan in 1999 and the related activities.


In addition, the Chinese-English translations in Section 2a are fed directly into the English-Korean MT software to provide four mappings between Chinese-Korean via transitive translation. The output for example query #55 used in Section 2a are shown below:
2c Web-based Entity Translation

Our assumption is that MT software can provide reasonably translation for general language expressions but may not be sufficient for entities such as names of person or places, etc. We implemented a web-based translation from English to Korean (and Chinese) that is oriented to entity names and terminology [5]. It is based on the normal convention of writers to express translations in bilingual document fragments in the following form: ... kkkkk (eeeeee) or ..eeeee (kkkkk), where kkkkk and eeeee are Korean and English strings respectively. When either of such patterns is encountered, it is quite likely that kkkkk will contain some kind of translation of eeeee, or vice versa. We search the web using an English term as key and request output snippets in Korean. These snippets are searched for the pattern above, and candidates for translation are isolated after some text processing and noise filtering.

This procedure was employed in E-K CLIR to translate any English terms that remain after the two MT software operations. Examples of these translations are also shown in Section 2b. Consider qry#2 via Systran: translations for ‘Jonnie’ were picked up with the indicated occurrence frequency in the returned web snippets. This was not performed for qry#2 via LniSoft because additional English words are adjacent to ‘Jonnie’. Our procedure regards such a word sequence as an indivisible phrase to gain precision, and try to locate its translation on the web. Apparently it failed. In qry#55, the translation for ‘Daepodong’ was also not found by our procedure.

3 Korean Text Processing and Indexing

Korean text is written with blank space as delimiter, but the characters in between can denote words, compounds or phrases [6]. For all tasks involving Korean, we employed a simple strategy of overlapping bigram indexing on the original texts without stemming or stopword removal as a default. In addition, we used a program called HAM version 6.0.0 [7] for the E-K and C-K retrieval tasks. HAM is an acronym for Hangul Analysis Module (or Model). It is a Korean lexical analyzer for Hangul (Korean) text. It supports an 'index' program which removes suffixes and stopwords and extracts simple nouns from compounds. For our indexing purposes, we keep both the simple nouns, the original compounds and the stemmed verbs, etc. We kept compounds because we can have some phrase indexing and also like to hedge concerning the outcome of segmentation. We call this HAM indexing.

4 Retrieval with Korean Collections:
4a K-K Monolingual Retrieval

Eight submissions using the Korean collection as retrieval target were submitted. Three were monolingual using title (T), description (D) and description with narrative (DN) sections of a topic to
form Korean queries. These serve as basis for evaluating other cross language retrievals with the Korean documents. Table 1 shows their results for the measures: R% (percent recall after 1000 retrieved), MAP (mean average precision), P10, P20 (average precision-at-10 and 20 documents retrieved) and R.Pre (average precision at the exact number of relevant documents for a query). Values of submitted runs are bolded in all tables. Rows with a * denote un-submitted runs. A 'b', 'w' or 'bw' following a run id denotes bigram, HAM indexing, or combination of these two retrieval lists.

Table 1 shows that monolingual Korean results have good MAP values (> 0.4) except in the case of D queries using rigid assessment (.3777). These queries are probably comparatively easy for the target collection. Average precision-at-10 for relax judgment range from 0.5561 to 0.6801. Queries of long (DN) type have better performance followed by short title (T) queries. D queries surprisingly perform some 18% worse than T type (MAP .4049 vs .4934, and the improvement is significant at the 5% level using sign test).

One general observation for the majority of our submitted runs is that D queries have worse MAP values than T queries for both Korean and English collections. This may be due to the fact that the short titles (of topic) have specific words and phrases only (e.g. qry#24: Illegal Tapping, Violation, Privacy), while the descriptions (of topics) are grammatical sentences often with only functional words added (qry#24: searching for documents dealing with the violation of people's privacy due to illegal tapping.)

Our submitted monolingual Korean retrieval makes use of bigram representation only. Table 1 shows also post-relevance-judgment runs using HAM indexing (stemming and stop-word removal) listed as: *pircs-K-K-T-01w and *pircs-K-K-D-02w. They are inferior to simple bigram indexing. Adding the bigram and word retrieval lists result in runs indicated by tag bw. They are not much different from the bigram only results.

4b E-K Crosslingual Retrieval

Table 2 shows results of English-Korean cross language retrieval. As discussed in Section 2, an English query was translated to Korean by both Systran and EnGuide. No pre-translation query expansion was employed, unlike [8]. Output from both was concatenated into a single query. This further went through our web-based translation to minimize the number of un-translated English terms. The resultant queries were indexed in two ways: directly via bigrams (b); and via stems produced by HAM (w). This would allow us to compare HAM indexing with bigrams. Our submissions pircs-E-K-T-01 and pircs-E-K-D-02 are however combination of retrieval lists from the two indexing schemes.
Table 2 shows that for E-K the MAP difference between T and D queries are small (.3598 vs .3566 relax) unlike K-K monolingual. Worth noting is that the precisions at 10 and 20 for D queries are about 10% better than for T (e.g. .5123 vs .4614). Apparently translation of the longer English D queries behaves similarly to T queries and can lead to translations more suitable for low-recall retrieval.

The E-K MAP values appear to be the best achieved among submissions. Compared to K-K monolingual retrieval, these crosslingual precision values attained 73% (T:.3598 vs .4934) and 82% (D:.3566 vs .4049) of relax effectiveness. The same comparisons for rigid assessment give: 78% (.3331 vs .4588) and 86% (.3249 vs .3777) respectively.

The un-submitted D run tagged ‘no web’ in Table 2 means no web entity translation was performed and can be compared with the submitted D run. This process has led to over 15% improvement (0.3064 vs. 0.3566 relax, significant at 5% level using sign test).

The * rows tagged ‘b’ and ‘w’ in Table 2 show un-submitted results of using bigram and HAM indexing scheme alone. The latter returns slightly worse MAP values than pure bigram: 0.3342 vs 0.3578 for T queries and 0.3154 vs 0.3388 for D queries. Combination of the two retrieval lists (our submitted results) however improves over both individually, unlike K-K runs.

In Table 2, we also show two bigram D runs that use either Systran (pircs-E-K-D-b-sys) or EnGuide (pircs-E-K-D-b-sys) translations only. These results are inferior (e.g., MAP for Systran is 0.2958, for EnGuide is 0.2581, compared to 0.3388 for pircs-E-K-D-02b where both translations were concatenated. Sign tests are significant at the 5% level for these improvements). This appears to support our assumption that MT combination leads to better effectiveness compared to using them singly.

We investigated why EnGuide results are inferior to Systran for description queries using bigram indexing. Part of the reason seems to be that entity names (like query #2: ‘Jonnie Walker Charity Golf Tournament’) in English queries are capitalized, and EnGuide has problem with them. Systran however is more flexible in regard to Ascii case and often provides the correct translation.

4c C-K Crosslingual Retrieval via English as Pivot

Results of our C-K retrieval using C\textsuperscript{C-K} transitive translation queries are tabulated in Table 3. Here, the Chinese queries (T and D only) first underwent a pre-translation expansion using the Chinese collections. (For DN queries, we assume they are sufficiently long that pre-translation would not have much effect.) We employ the top 10 documents of an initial retrieval and added conservatively only 6 terms to each query. The queries were translated two ways into English using Systran and Loto packages as discussed in Section 2. The English output were further translated into Korean by Systran and EnGuide, resulting in four Korean mappings for each query. Any English terms left un-translated were processed by our web-based translation. The final queries were then indexed two ways bigram (b) and HAM indexing (w) as in E-K. The submitted results use combination of retrieval lists from (b) and (w) runs.

An error was later discovered in the PRF process for the description pircs-C-K-D-02 run, which is tagged with ‘e’. (The number of feedback documents was random for this run.) The next row *C-K-D-02 bw without error tag ‘e’ tabulates the corrected values. It is about 5-6% better.

The C-K relax assessment MAP values range between 0.2784 D queries to 0.3076 for DN queries. They represent 56% (T queries: .2783/.4934), 69% (D: .2784/4049) and 60% (DN: .3076/.5161) of monolingual K-K retrieval. For rigid assessment, these ratios are: 56%, 70% and 61% respectively. Short title queries have worst comparison to K-K monolingual.

| C-K-D-02 bw e | 69 | .2601 | .3895 | .3518 | .2855 |
| C-K-D-02 bw | 71 | .2784 | .3965 | .3658 | .2923 |
| C-K-D-02 b | 62 | .2402 | .3123 | .2965 | .2640 |
| C-K-D-02 w | 73 | .2718 | .3930 | .3561 | .2908 |
| C-K-D-02 bsys | 69 | .2681 | .3807 | .3447 | .2905 |
| C-K-DN-03 bw e | 70 | .3076 | .4281 | .3737 | .3181 |

a) Relax Assessment (number relevant = 3917)

| C-K-D-02 bw e | 70 | .2471 | .3526 | .3202 | .2706 |
| C-K-D-02 bw | 73 | .2632 | .3596 | .3298 | .2782 |
| C-K-D-02 b | 63 | .2260 | .2807 | .2632 | .2513 |
| C-K-D-02 w | 74 | .2555 | .3509 | .3219 | .2734 |
| C-K-D-02 bsys | 70 | .2518 | .3386 | .3061 | .2778 |

b) Rigid Assessment (number relevant = 3131)

Table 3a,b: C-K Cross Language Results for Query Types T, D, DN.
We would like to have direct Q<sup>C-K</sup> bilingual retrieval results for comparison purposes but could not find resources for the direct C-K translation. Table 3 also shows results using bigram indexing or word indexing alone. Here, bigram indexing returns much worse than HAM indexing. It seems that going through four translations lead to proportionately more suffixes than content. Meaningless bigrams proliferate and becomes a factor. With HAM processing, stems and stopwords are removed and we do not have that much noise. The two rows tagged with ‘bsys’ are bigram runs with queries that concatenate only two Systran English-Korean translations (without EnGuide). Their results improve to close the gap with HAM indexing results.

5 Retrieval with English Collections:

5a E-E Monolingual Retrieval

English monolingual retrieval was performed to provide a basis for evaluating our Chinese-English crosslingual retrieval. Results are tabulated in Table 4. We employed Porter’s stemming, stopword removal, and PRF procedures in our runs.

<table>
<thead>
<tr>
<th>pircs-</th>
<th>R%</th>
<th>MAP</th>
<th>P10</th>
<th>P20</th>
<th>R.Pre</th>
</tr>
</thead>
<tbody>
<tr>
<td>E-E-T-01</td>
<td>62</td>
<td>.2879</td>
<td>.5017</td>
<td>.4888</td>
<td>.3319</td>
</tr>
<tr>
<td>*E-E-T-01 x-</td>
<td>62</td>
<td>.3235</td>
<td>.5069</td>
<td>.4888</td>
<td>.3494</td>
</tr>
</tbody>
</table>

a) Relax Assessment (number relevant = 11056)

<table>
<thead>
<tr>
<th>pircs-</th>
<th>R%</th>
<th>MAP</th>
<th>P10</th>
<th>P20</th>
<th>R.Pre</th>
</tr>
</thead>
<tbody>
<tr>
<td>E-E-D-01</td>
<td>62</td>
<td>.3876</td>
<td>.5845</td>
<td>.5638</td>
<td>.3998</td>
</tr>
</tbody>
</table>

b) Rigid Assessment (number relevant = 5866)

Table 4a,b: E-E Monolingual Results for 58 Query Types T, D.

5b C-E Crosslingual Retrieval

Chinese-English bilingual retrieval was done as an intermediate step to our goal of C-K pivot retrieval. The process has been discussed in Section 4c. These results are tabulated in Table 5. The relax assessment MAP values of 0.2879 for T and 0.2829 for D queries appear to be the top results among participants. These represent 71% (T) and 73% (D) compared to our E-E monolingual retrieval relax assessment (Table 4a), and 75% and 73% for rigid assessment.

The Q<sup>C-K</sup> queries do not have the assistance from web-assisted translation. The retrieval result supports our observation that the MT software are reasonably adequate for CLIR purposes.

Table 5 also shows two un-submitted runs that do not include pre-translation expansion (x-). To our surprise, MAP values without pre-translation are better than with pre-translation. Apparently the MT software themselves provide sufficiently good translations. Another two un-submitted D-query runs show results of using MT software individually: tagged as ‘sys’ and ‘lot’. Systran is better than Loto translation. Just as in E-K, the concatenated translation results for pircs-C-E-D-02 are better than these that use translations singly. Here however, the improvements are not significant according to the sign test at the 5% level.

<table>
<thead>
<tr>
<th>pircs-</th>
<th>R%</th>
<th>MAP</th>
<th>P10</th>
<th>P20</th>
<th>R.Pre</th>
</tr>
</thead>
<tbody>
<tr>
<td>C-E-T-01</td>
<td>68</td>
<td>.2380</td>
<td>.3862</td>
<td>.3707</td>
<td>.2746</td>
</tr>
<tr>
<td>*C-E-T-01 x-</td>
<td>63</td>
<td>.2471</td>
<td>.3586</td>
<td>.3293</td>
<td>.2692</td>
</tr>
</tbody>
</table>

b) Rigid Assessment (number relevant = 5866)

Table 5a,b: C-E Crosslingual Results for 58 Query Types T, D.

6 Chinese C-C Monolingual Retrieval

Chinese monolingual retrieval was performed as in last year: based on combination of retrieval lists using bigram and word indexing. Results are shown in Table 6; they provide a basis for CLIR involving Chinese collections.

<table>
<thead>
<tr>
<th>pircs-</th>
<th>R%</th>
<th>MAP</th>
<th>P10</th>
<th>P20</th>
<th>R.Pre</th>
</tr>
</thead>
<tbody>
<tr>
<td>C-C-T-01</td>
<td>84</td>
<td>.2673</td>
<td>.3373</td>
<td>.2864</td>
<td>.2725</td>
</tr>
<tr>
<td>C-C-D-02</td>
<td>86</td>
<td>.2761</td>
<td>.3542</td>
<td>.2941</td>
<td>.2810</td>
</tr>
</tbody>
</table>

b) Rigid Assessment (number relevant = 1318)

Table 6a,b: C-C Monolingual Results for 59 Query Types T, D.

7 Conclusion and Discussion
We tested several MT packages for cross language retrieval purposes: Chinese to English, English to Korean. These are augmented with a web-based entity/terminology-oriented translation procedure. Experiments show that concatenation of two translations performs better than using them singly for direct C-E and E-K CLIR. Individually, Systran translation for C-E has better retrieval outcome than Loto, and for E-K Systran is better than EnGuide. These CLIR runs provide 71% to 88% of monolingual effectiveness for direct bilingual retrieval operations.

Our web-based entity/terminology-oriented translation is found effective, and can provide some 15% improvement in mean average precision.

In Korean retrieval, bigram provides better effectiveness than word indexing except in C-K runs where a query has a combination of 4 translations and random bigram noise may become an adverse factor. In general, combination of their retrieval lists provide better effectiveness except for K-K title run.

Fig.7 summarizes our results compared to the official Max and Median of all submitted runs from participants.

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**References**