

A Method of Cross-Lingual Question-Answering
Based on Machine Translation and Noun Phrase
Translation using Web documents

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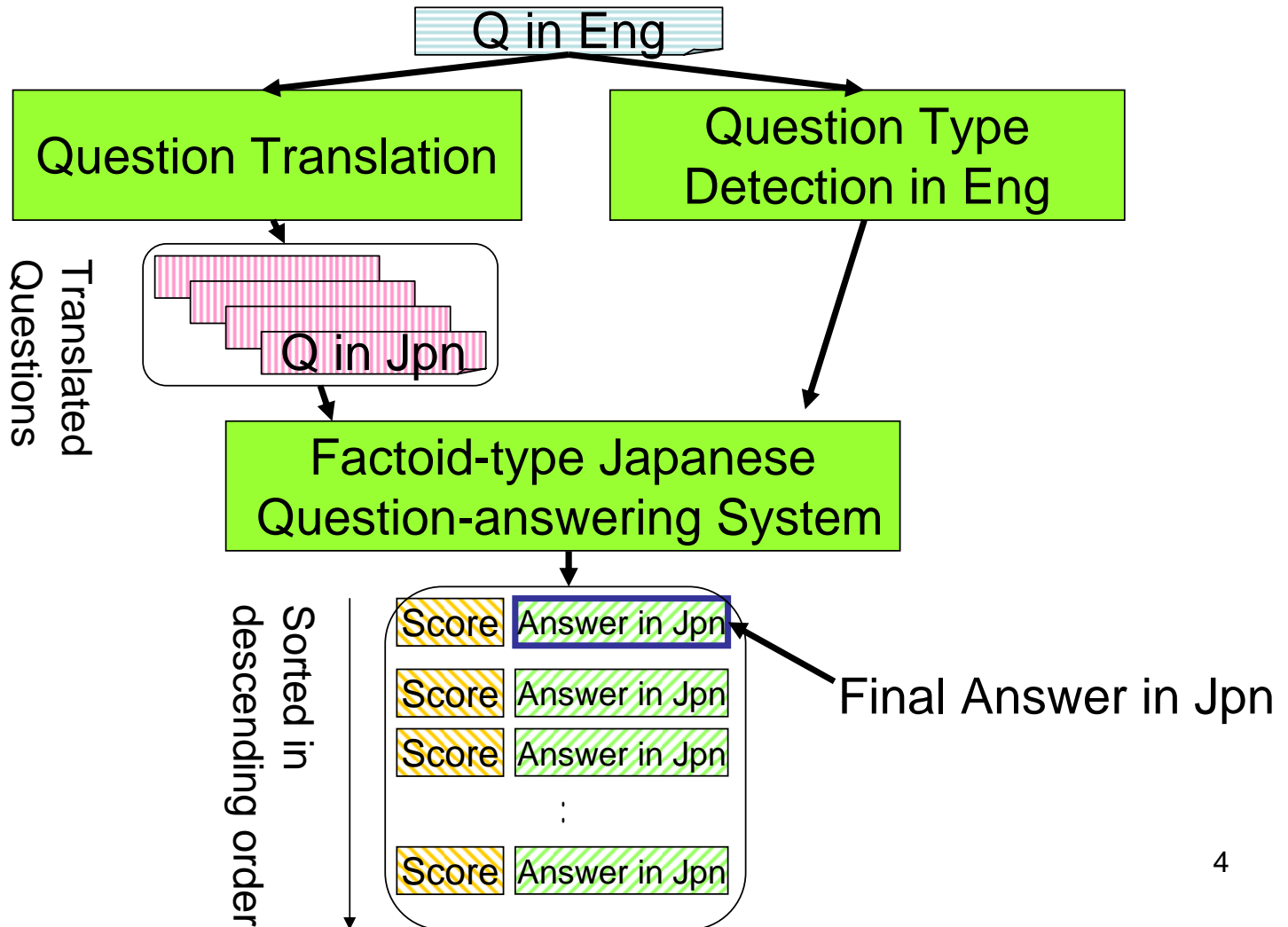
Introduction and related work

- Cross-lingual Question Answering
 - a. For each target language, one individual QA system is prepared. The CL process is achieved as the translation of Qs.
 - b. One pivot language is assumed and one QA system is prepared. The CL process appears in the translation of Qs and/or documents.
- While some researches adopt the second approach [Bowden 06, Laurent 06, Shimizu 05, Mori 05], the majority adopts the first approach.
- One of main concerns is the improvement of translation accuracy.
- Web as resource to translate Out-of-Vocabulary (OOV) words
 - Zhan et al. [Zhang 05] proposed a method to obtain translation candidates from the results of a search engine.
 - Bouma et al. [Bouma 06] extracted from English Wikipedia all pairs of lemma titles and cross-links to the corresponding link to Dutch Wikipedia.

Our approach

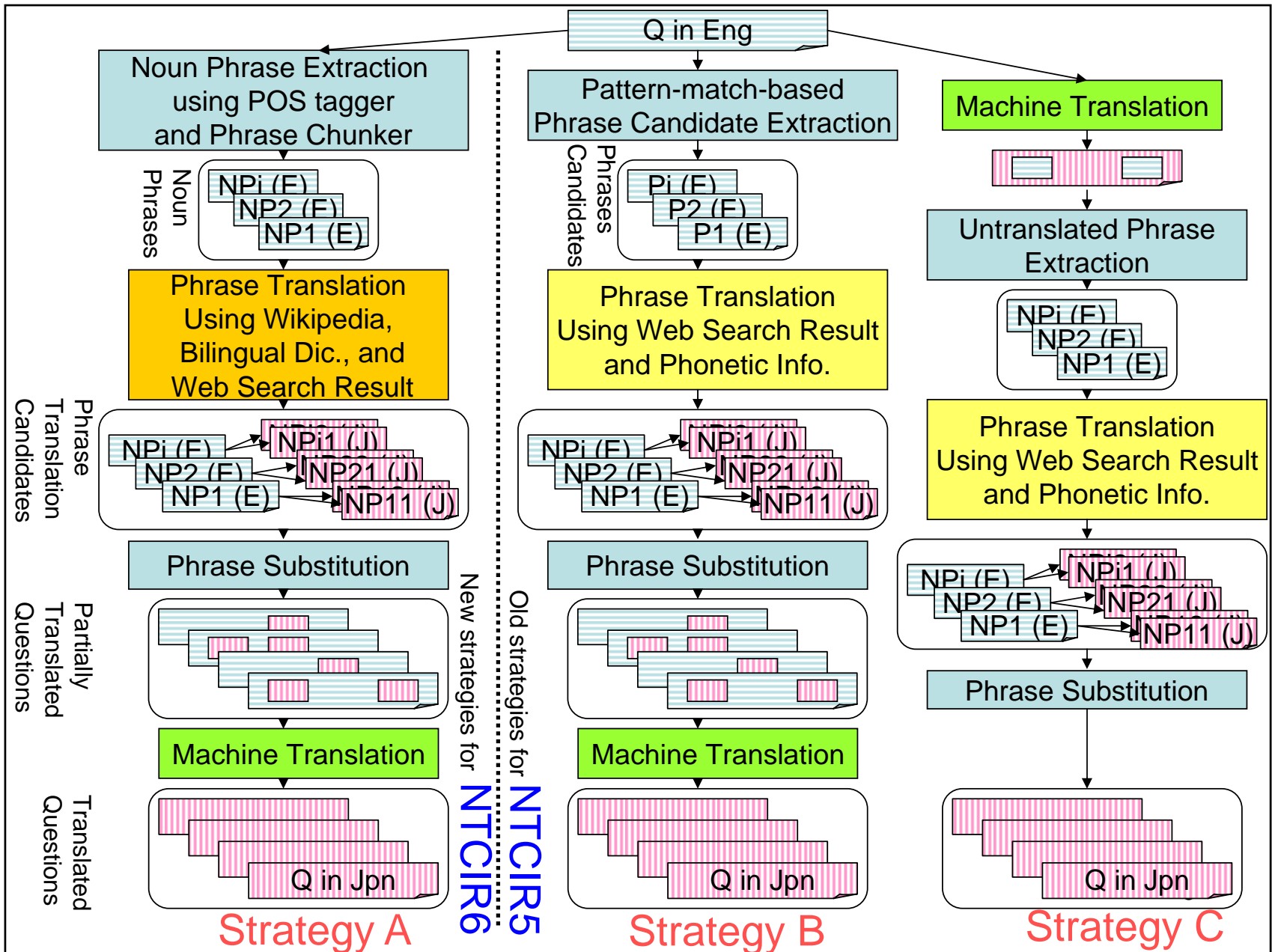
- English-Japanese CLQA
- A question translation approach (next slide)
 1. Translate an English Q. into Japanese
 2. Detect the Q. type in the English Q.
 3. Perform Japanese QA with translated Qs.
- Points at issues
 - Treatment of OOV phrases in combination with MT
 - Many off-the-shelf MT products are available.
 - Translation of English Q. into Japanese by using MT.
 - Out-of-vocabulary (OOV) phrases
 - Management of multiple translation candidates in QA phase
 - Different translation strategies of OOV phrases yield different translated Q.

A question translation approach



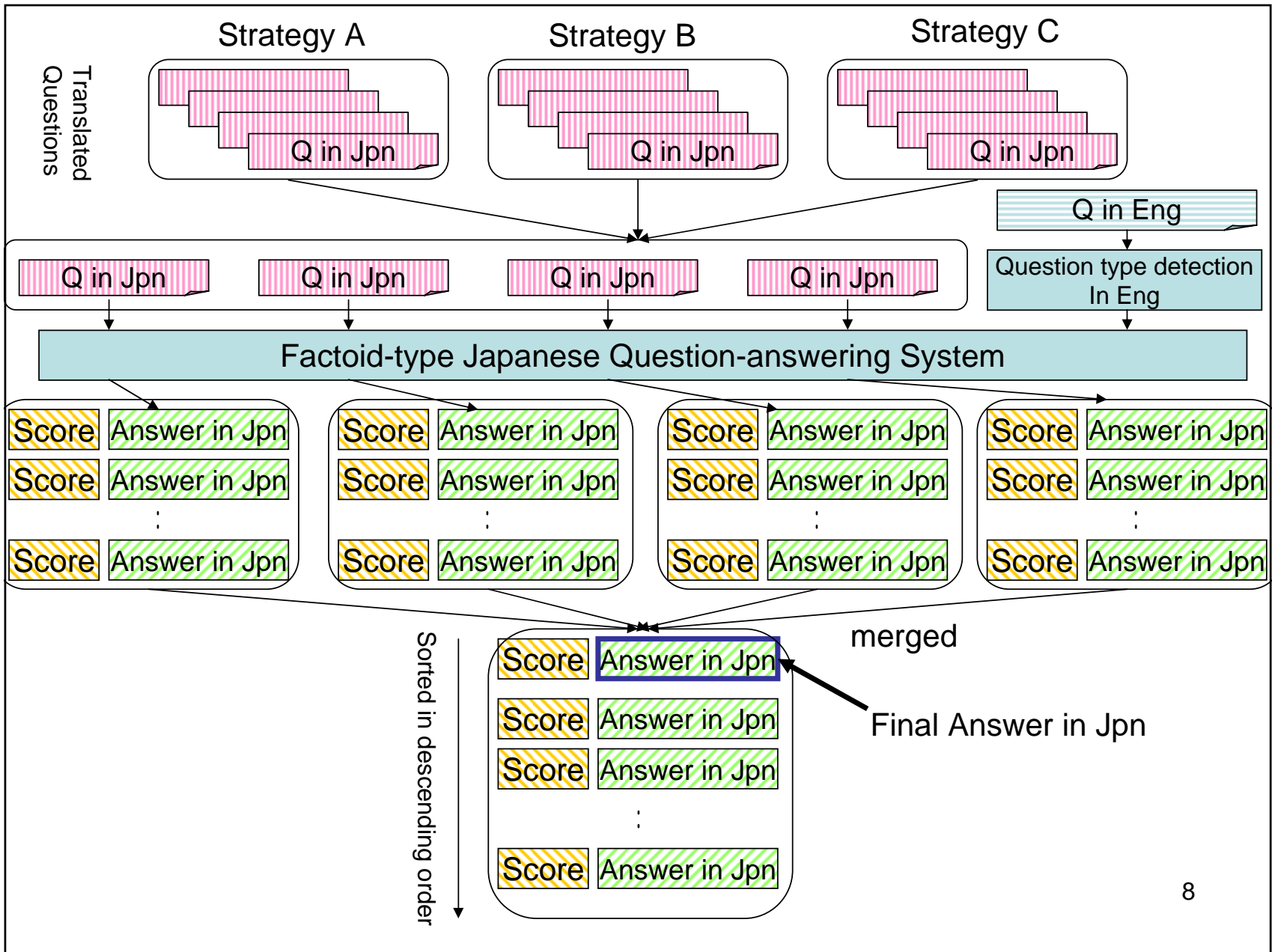
Treatment of OOV phrase in combination with an MT

- Translation of OOV phrases using external resource
 - There are several different approaches that are worth employing (described later)
- Timing of combining translation of OOV phrases with an MT
 - As a pre-editing process of MT
 - Some of E-J MT systems can treat Japanese strings in an input English sentence as unknown noun phrases and outputs them as they are.
 - *Pre-translation*: originally a technique to utilize Translation Memory
 - Partial translation of noun phrases first, then perform MT
 - As a post-editing process of MT
 - MT first, then translate un-translated noun phrases.
 - We do not have ways to correct translation error in MT.



Management of multiple translation candidates in QA phase

- Multiple translation candidates of Q. from different translation strategies
 - Which is the best translation? → No criterion
- “Cohesion with information source” approach.
 - Hypothesis 1: if the translation is performed well, some context similar to the translated Q. is likely found in information source.
 - “Answering a question” is finding objects whose context in the information source is coherent with the question.
 - Hypothesis 2: the degree of cohesion with information source is analogous to the appropriateness of the answer candidate.
 - E.g. Score of answer



Translation strategies

- Strategy A: newly introduced for NTCIR-6 CLQA
 - Performed as the pre-translation process.
 - SVM-based NP chunker to extract all possible NPs.
 - Phrase translation using Wikipedia
 - Phrase translation using Web search results
- Strategy B and C: introduced for NTCIR-5 CLQA
 - Translate loan words into the original Japanese words using Web and the information of pronunciation.
 - B is performed as the pre-translation process.
 - C is performed as the post-translation process.

Phrase translation using Wikipedia

- Wikipedia is a free content encyclopedia, and has a lot of articles in more than 200 languages.
- We can easily obtain multilingual translation of an entry term because of hyper-links [Bouma 06, Fukuhara 07].
 1. To perform the E-J translation, search for **the target phrase in the English Wikipedia**.
 2. Find out the **link to the corresponding Japanese entry**.
 3. **The name of the Japanese entry** is expected to be a proper translation.
- We may use not only English entries but also other entries in different languages that have similar alphabets.

Algorithm 4.1: WIKIPEDIATRANSEJ(*PhraseE*)

```
for each langCode ∈ {‘en’, ‘de’, ‘fr’, ‘es’, ‘it’,  
                    ‘nl’, ‘pt’, ‘lt’, ‘no’, ‘pl’}  
do { art ← Wikipedia(PhraseE, langCode)  
    if (art has a link to a Japanese entry PhraseJ)  
    then return (PhraseJ)  
return (“)
```

Phrase translation using Web search results (1)

- We propose a modification of Zhang's method [Zhang 05].
- Main idea: the case of E-J translation
 - Submit an **English phrase** to a Web search engine in order to retrieve **Japanese documents**.
 - Many of retrieved documents are expected to contain not only the English phrase but also **Japanese phrases** that related to the original English phrase.
 - Scoring method that estimate the appropriateness of the candidate in terms of translation.

Phrase translation using Web search results (2)

Algorithm 4.2: WEBTRANSCAND(*Phrase*)

main

$\langle T, S \rangle \leftarrow \text{CallExternalEngine}(\text{Phrase}, N_d)$

comment: T and S are the arrays of titles and snippets, respectively. The index corresponds to the rank in the search result. N_d is the number of document to be retrieved.

$TC \leftarrow \{\}$

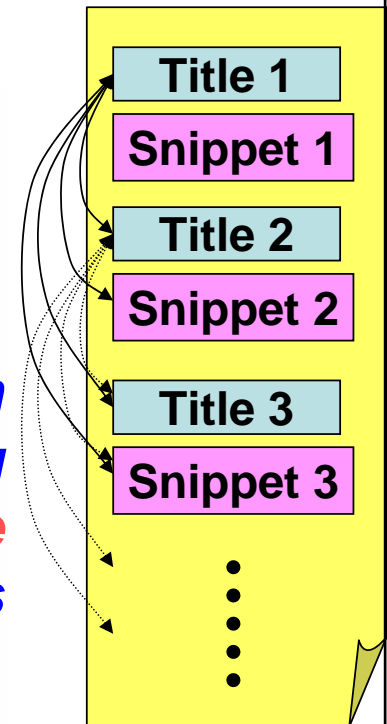
for $i \leftarrow 1$ **to** $n - 1$

do $\left\{ \begin{array}{l} \text{for } j \leftarrow i + 1 \text{ to } n \\ \text{do } \left\{ \begin{array}{l} LCSS_T \leftarrow LCCS(T_i, T_j) \\ LCSS_S \leftarrow LCCS(T_i, S_j) \\ TC \leftarrow \text{MergeFreq}(TC, LCSS_T) \\ TC \leftarrow \text{MergeFreq}(TC, LCSS_S) \end{array} \right. \end{array} \right.$

Candidates: Longest Common Contiguous Substring of Japanese characters

return (TC)

comment: $TC = \{\langle C_1, freq_1 \rangle, \langle C_2, freq_2 \rangle, \dots\}$, where C_i and $freq_i$ are a translation candidate of *Phrase* and its frequency, respectively.



Search Result

Phrase translation using Web search results (3)

- Assigning score to each candidate

- Zhang's original score

- $ITF(C_i)$: Inverse of translation freq. that represents how many times the translation candidate C_i appears in different candidate lists.

$$Scorg(C_i) = \alpha \cdot \frac{freq_i}{maxFreq(TC)} \cdot ITF(C_i) + (1 - \alpha) \cdot \frac{1}{Rank(C_i) + \beta}$$

ITF is properly calculated only when we want to translate **a number of phrases simultaneously**.

- Our modification

$$Screv(C_i) = \left(\alpha \cdot \frac{\log_2(freq_i)}{\log_2(maxFreq(TC))} + (1 - \alpha) \cdot \frac{1 + \beta}{Rank(C_i) + \beta} \right)$$

Since the algorithm tends to produce shorter candidate, we give **“reward” to longer one**.

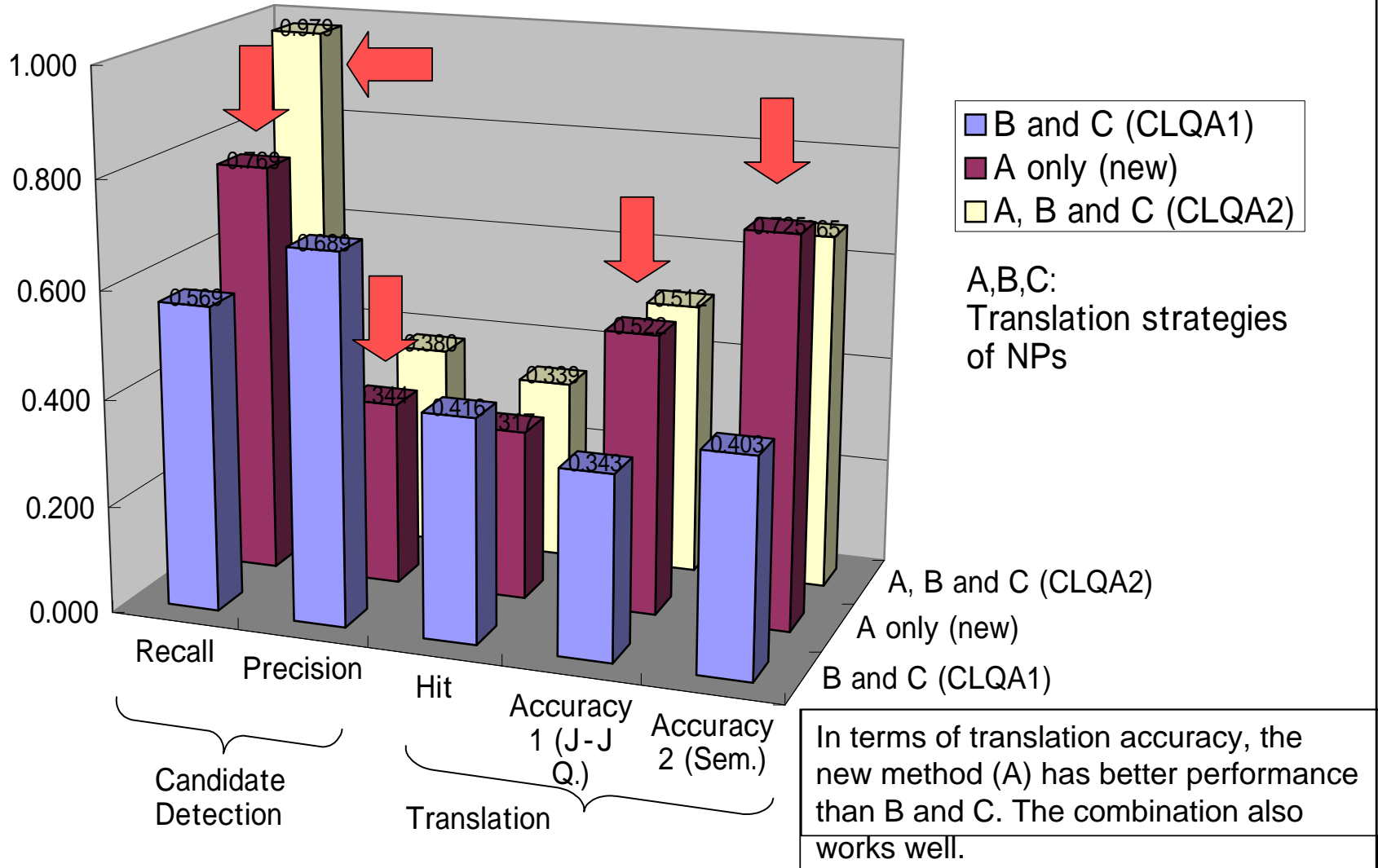
Runs at NTCIR-6 CLQA

- Participated in the English-Japanese task.
- Settings
 - An off-the-shelf MT product that has “pre-translation” function (IBM Japan, *Hon'yaku-no Ousama*)
 - EDR E-J translation dictionary
 - A Japanese QA system for factoid Qs. [Mori 05]
 - Strategy A
 - Web search engine: Web service by Yahoo! Japan
 - Strategy B and C
 - The setting is same as our formal run in NTCIR-5 CLQA.
 - Web search engine: Google SOAP Search API.
- Runs
 - Forst-E-J-01: Strategy A, B, and C with MT
 - Forst-E-J-02: Strategy A with MT
 - Forst-E-J-03: Strategy B and C with MT (NTCIR-5 CLQA)
 - Forst-J-J-01: Mono-lingual run. An upper bound.
 - Baseline: MT only

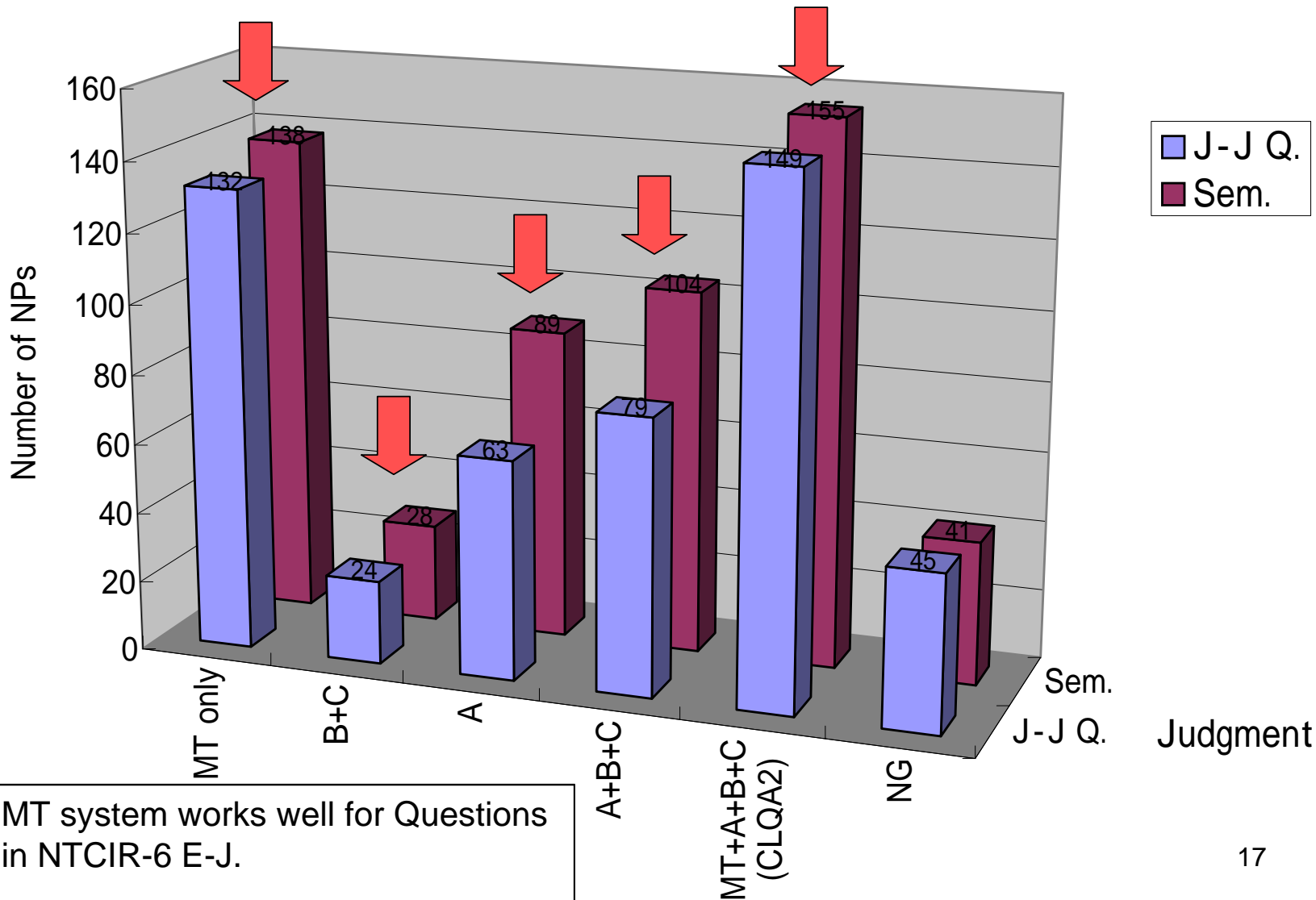
Performance of proper noun translation

- Measures for evaluation of proper noun detection
 - Recall and precision
- Measures for evaluation of proper noun translation
 - **Hit**: ratio of # of phrases to which the system can find at least one translation candidate.
 - **Trans. Accuracy 1**: ratio of # of phrases for which the system can find at least one “correct” translation. “correct” when the translation is the correspondent phrase in J-J Q. (strict)
 - **Trans. Accuracy 2**: same as **1**, but the correctness is judged semantically. (lenient)

Performance in translation of proper NPs

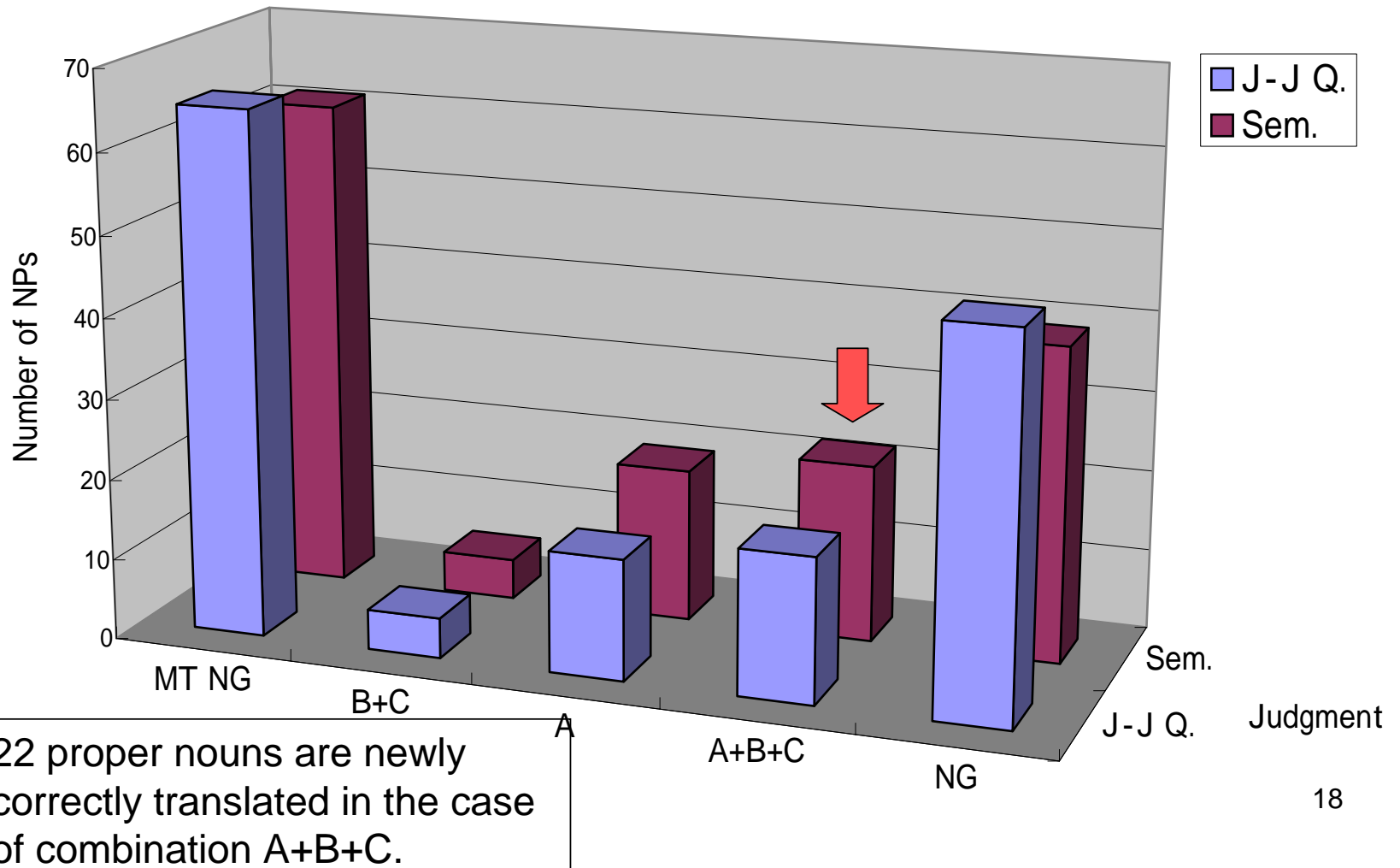


Number of correctly translated proper NPs



MT system works well for Questions in NTCIR-6 E-J.

Number of correctly translated proper NPs which the MT cannot translate



Performance in E-J CLQA

Strategy	Run ID	Acc	MRR	TOP5	Acc+ U	MRR +U	TOP5 +U
MT only		0.175	0.195	0.230	0.185	0.230	0.315
MT+B+C (CLQA1)	Forst-E-J-03	0.170	0.193	0.235	0.180	0.229	0.325
MT+A	Forst-E-J-02	0.170	0.192	0.230	0.180	0.231	0.325
MT+A+B+C	Forst-E-J-01	0.175	0.197	0.230	0.195	0.244	0.320
JJ QA	Forst-J-J-01	0.310	0.361	0.440	0.335	0.410	0.525

Acc: Accuracy

+U: Unsupported answers are allowed

JJ QA: Japanese monolingual QA system with correct Japanese questions.

- Although “MT+A+B+C” has better performance than others, the difference between it and “MT only” is not significant.
- MT system works well and the actual improvement by phrase translation is small.

Failure in extracting NPs.

- Adjacent proper nouns are extracted as one phrase
 - **Question:** “Where did former Spice Girl Posh Spice hold her wedding ceremony?”
 - **Extracted NP:** “Spice Girl Posh Spice”
 - **Correct NPs:** “Spice Girl” and “Posh Spice”

Failure in phrase translation by using Wikipedia

- Translation using Wikipedia mostly works well, when it is applicable.
- It has unwilling tendency to translate a NP into an official name of translation instead of a popular translation.
 - **Phrase:** “Akutagawa Prize”
 - **Translated:** “akutagawa ryunosuke shou” (芥川龍之介賞)
 - **More popular translation:** “akutagawa shou”(芥川賞)

Failure in phrase translation by using Web search result

- The method tends to fail in translation of longer NPs.
 - NP: “University of Hawaii at Manoa”
 - Translated: “hawai daigaku” (ハワイ大学)
 - Correct one: “hawai daigaku manoa kou” (ハワイ大学マノア校)
- It also tends to translate a phrase into a related phrase.
 - NP: “FIFA president”
 - Translated: “sakkaa” (football, サッカー)
 - Correct one: “FIFA kaichou” (FIFA会長)

Concluding remarks

- English-Japanese (E-J) task with three systems.
 - Basis of approach: MT + an existing Japanese QA system.
 - Methods for noun phrase translation using the Web.
- The combination works well.
- MT system also works well for Qs in NTCIR-6 E-J.