Neural Reranking Improves Subjective Quality of Machine Translation: NAIST at WAT 2015

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Statistical Translation Frameworks

Symbolic Models





Relative Merits/Demerits

- Symbolic Models
 - Inner workings well understood
 - Better at translating low-frequency words
- Continuous-space Models
 - Easier to implement
 - Produce more fluent output
 - Probabilistic model can score output of other systems!



Reranking with Neural MT Models





What Do We Know About Reranking?

 Reranking greatly improves BLEU score, even over strong baseline systems:



<u>Alkhouli+ 2015</u>

	de-en BLEU	ar-en BLEU
Baseline	30.6	26.4
Reranked	32.3	27.0



What **Don't** We Know About Reranking?

- Does reranking improve subjective impressions of results?
- What are the qualitative differences before/after reranking with neural MT models?



Experiments



Experimental Setup

- Data: ASPEC Scientific Abstracts
 - Japanese \leftrightarrow English, Chinese
- Baseline: NAIST WAT2014 Tree-to-String System
 - Strong baseline achieving high scores
 - Implemented using Travatar (<u>http://phontron.com/travatar</u>)
- Neural MT Model: Attentional model
 - Trained ~500k sent., 256 hidden nodes, 2 model ensemble
 - Use words occurring 3+ times (vocab 50,000~80,000)
 - Trained w/ lamtram (http://github.com/neubig/lamtram)
- Automatic Evaluation: BLEU, RIBES
- Manual Evaluation: WAT 2015 HUMAN Score



Confirm what we know: Neural reranking helps automatic evaluation



Show what we didn't know: Also help manual evaluation.

What is Getting Better?

- Perform detailed categorization of the changes in Japanese-English results:
 - 1. Is the sentence better/worse after ranking?
 - 2. What is the main error corrected: insertion, deletion, substitution, reordering, or conjugation?
 - 3. What is the detailed subcategory?



Main Types of Errors Corrected/Caused

Туре	Improved	Degraded	% Impr.	
Reordering	55	9	86%	}_
Deletion	20	10	67%	
Insertion	19	2	90%	
Substitution	15	11	58%	
Conjugation	8	1	89%	$\mathcal{F} \setminus \mathcal{F}$
Total	117	33	78%	
Overall improvements re-confirmed				

In particular fixing reordering, insertion, and conjugation errors



#1 Detailed Improvement Category: Phrasal Reordering (+26, -4)

<u>Source</u>

症例2においては、直腸がんの肝転移に対する化学療法中に、 発赤、硬結、皮膚潰ようを生じた。

<u>Ref</u>

In case 2, reddening, induration, and skin ulcer appeared during chemical therapy for liver metastasis of rectal cancer.

<u>Base</u>

In case 2, occurred during chemotherapy for liver metastasis of rectal cancer, flare, induration, skin ulcer.

<u>Rerank</u>

In case 2, the flare, induration, skin ulcer was produced during the chemotherapy for hepatic metastasis of rectal cancer.

#2 Detailed Improvement Category: Auxiliary Verb Ins./Del. (+15, -0)

<u>Source</u>

これにより得られる支配方程式は壁面乱流のようなせん断乱流に も有用である。

<u>Ref</u>

Governing equation derived by this method is useful for turbulent shear flow like turbulent flow near wall.

<u>Base</u>

The governing equation **is** obtained by this is also useful for such as wall turbulence shear flow.

<u>Rerank</u>

The governing equation obtained by this is also useful for shear flow such as wall turbulence.

#3 Detailed Improvement Category: Coordinate Structures (+13, -2)

<u>Source</u>

レーザー加工は高密度光束による局所的な加熱とアブレーション により行う。

<u>Ref</u>

Laser work is done by local heating and ablation with high density light flux.

<u>Base</u>

The laser processing is carried out by local heating by highdensity luminous flux and ablation.

<u>Rerank</u>

The laser processing is carried out by local heating and ablation by high-density flux.



#4 Detailed Improvement Category: Verb Agreement (+6, 0)

Source

ラングミュア-ブロジェット法や包接化にも触れた。

<u>Ref</u>

Langmuir-Blodgett method and inclusion compounds are mentioned.

<u>Base</u>

Langmuir-Blodgett method and inclusion is also discussed.

<u>Rerank</u>

Langmuir-Blodgett method and inclusion are also mentioned.



What Wasn't Helped: Terminology (+2, -4)

<u>Source</u>

放射熱を利用する赤外線応用計測が応力解析に役立っている

<u>Ref</u>

Infrared ray applied measurement using radiant heat is useful for stress analysis.

<u>Base</u>

The infrared application measurement using radiant heat is useful in the stress analysis.

<u>Rerank</u>

Infrared ray application measurement using radiation heat is useful for stress analysis.



Conclusion



What Do We Know Now?

- Neural reranking improves subjective quality of machine translation output.
- Main gains are from grammatical factors, and not lexical selection.



What Do We Still Not Know Yet?

- How do neural translation models compare with neural language models?
- How does reranking compare with pure neural MT?



Thank You!