Translation reranking using source phrase dependency features Antonio Valerio Miceli-Barone Università di Pisa

Rerank N-best list of translations using source dependency syntax and phrase alignments



Phrase segmentation and one-to-one alignment generated by the phrase-based decoder Phrase dependencies inspired by Gimpel and Smith (2013)

Results

- ·Experiments on *it-en* Europarl v7+JRC-ACQUIS v2.2+additional
- \cdot Training set size: 3,075,777 s.p.
- \cdot Tuning set size: 3,923 s.p.
- Baseline: Moses with sparse features ("word translation" and "phrase translation" Chiang et al. (2009)) ·BLEU:
 - ·Moses + sparse feats:
 - Moses + sparse feats. + rerank: 29.17

• Unambiguous PARENTS, un. L. PARENTS, un. R. PARENTS • Unique PARENT • No CHILDREN, no L. CHILDREN, no R. CHILDREN One-sided CHILDREN Distortion: detect reordering. For each pair of source phrases adjacent in target order: • Unique parent-child / unique child-parent Siblings with unique parent None of the above All conjoined with an *inversion feature*.

 Train and tune standard phrase-based (Moses) system 1000-best decode tuning set Parse source sentences with DeSR (Attardi and Ciaramita 2007) For each source sentence and candidate translation: Compute phrase dependency features Concatenate with Moses feature vector • Re-tune Moses linear model on tuning set augmented by phrase dependency features K-best batch MIRA (Cherry and Foster, 2012) • Use this linear model to rerank candidate translations of new sentences

29.02 +0.15 Features

Segmentation: detect syntactic subtree breakage by the decoder • No PARENTS, no LEFT PARENTS, no RIGHT PARENTS

System