

SIMULATING SENTENCE PAIRS SAMPLING VIA SOURCE AND TARGET LANGUAGE MODELS

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Outline



- Motivations
- Proposed Approach
- Experiments



Motivations

Weighting Sentence Pairs

- Normal word alignment

- Each sentence pair (e^k, f^k) is assigned an empirical probability $\hat{P}(e^k, f^k)$
- IBM Model 1: lexicon probability of source word \mathbf{f} given target \mathbf{e}

$$p(\mathbf{f}|\mathbf{e}) = \frac{\sum_k c(\mathbf{f}|\mathbf{e}; e^k, f^k)}{\sum_{k, \mathbf{f}} c(\mathbf{f}|\mathbf{e}; e^k, f^k)} \quad (1)$$

$$c(\mathbf{f}|\mathbf{e}; e^k, f^k) = \sum_{e^k, f^k} \hat{P}(e^k, f^k) \sum_a P(a|e^k, f^k) \cdot (2)$$
$$\sum_j \delta(\mathbf{f}, f_j^k) \delta(\mathbf{e}, e_{a_j}^k)$$

- $\hat{P}(e^k, f^k)$ is estimated by MLE on the full sentence pairs which would give most uniform probabilities ($\sim 1/S$)

Motivation

- It's helpful if $\hat{P}(e^k, f^k)$ can approximate the true distribution $P(e^k, f^k)$
- $\hat{P}(e^k, f^k)$ is a prior
- Some sentences could be more valuable, reliable, appropriate, and should therefore have a **higher weight** in the training
- Can we have better a approximation for $\hat{P}(e^k, f^k)$?



Proposed Approach

Proposed approach

- $\hat{P}(e^k, f^k) \sim$ sentence pair confidence (**sc**)
 - ▣ Quality of sentence pair for training the alignment model
- $\hat{P}(e^k, f^k) \sim$ genre-dependent sentence pair confidence (**gdsc**)
 - ▣ Appropriateness of a sentence pair to train a system for a specific genre
- Sentence-dependent phrase alignment confidence (**sdpc**) scores
 - ▣ Which sentence pairs the phrase pair was extracted

Sentence pair confidence (sc)

- It's hard to compute $P(e^k, f^k)$ without knowing $P(e^k | f^k)$ which is estimated during the alignment process

- Assumption

$$\hat{P}(e^k, f^k) = P(e^k)P(f^k)$$

- $P(e^k)$, $P(f^k)$ can be estimated by source and target language models

Sentence pair confidence (sc)

- Average log likelihood of each sentence pair

$$\begin{aligned}\mathcal{L}(e^k) &= \frac{\sum_{e_i^k \in e^k} \log P(e_i^k | h)}{|e^k|} \\ \mathcal{L}(f^k) &= \frac{\sum_{f_j^k \in f^k} \log P(f_j^k | h)}{|f^k|}\end{aligned}\quad (3)$$

$$\mathcal{L}(e^k, f^k) = [\mathcal{L}(e^k) + \mathcal{L}(f^k)]/2$$

- Sentence pair confidence score (sc)

$$\begin{aligned}sc(e^k, f^k) &= \exp(\mathcal{L}(e^k, f^k)) \\ &= \sqrt{\left(\prod_{e_i^k \in e^k} P(e_i^k | h)\right)^{-|e^k|} \left(\prod_{f_j^k \in f^k} P(f_j^k | h)\right)^{-|f^k|}}\end{aligned}\quad (4)$$

Genre-dependent sentence pair confidence (gdsc)

- Adopt training data toward a target genre.
- Use genre-dependent language models to assign sentence pair confidence

- Given genre ***g***

$$gdsc(e^k, f^k) = sc(e^k, f^k | g) \quad (5)$$

- Average likelihood of each sentence is estimated by genre-specific language models

Sentence-dependent phrase alignment confidence (sdpc)

- We want to put sc into decoding process
 - ▣ Add a feature in phrase pairs
- Track from which sentence pairs the phrase pair was extracted
- Given a phrase pair (ep, fp) , the **sdpc** score

$$sdpc(ep, fp) = \exp \frac{\sum_{(e^k, f^k) \in S(ep, fp)} \log sc(e^k, f^k)}{|S(ep, fp)|}$$
$$S(ep, fp) = \{(e^k, f^k) | ep \in e^k, fp \in f^k\} \quad (6)$$

where $S(ep, fp)$ is the set of sentences that the phrase pair come from



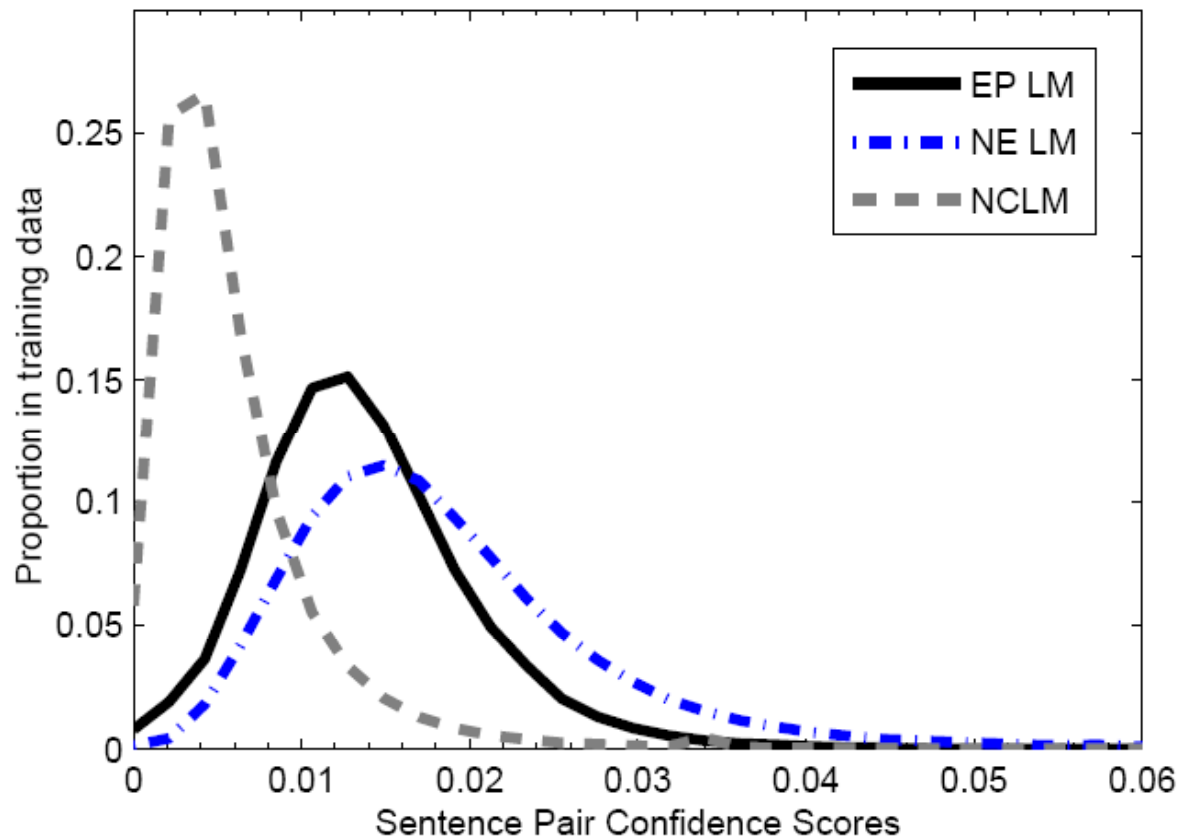
Experimental Results

Set-up

- EN ↔ ES
- Training & test data from 2 genres
Europarl and News-Commentary (ACL'08-WMT)
- Standard toolkits
Moses, SRILM, GIZA++ (multi-threaded)

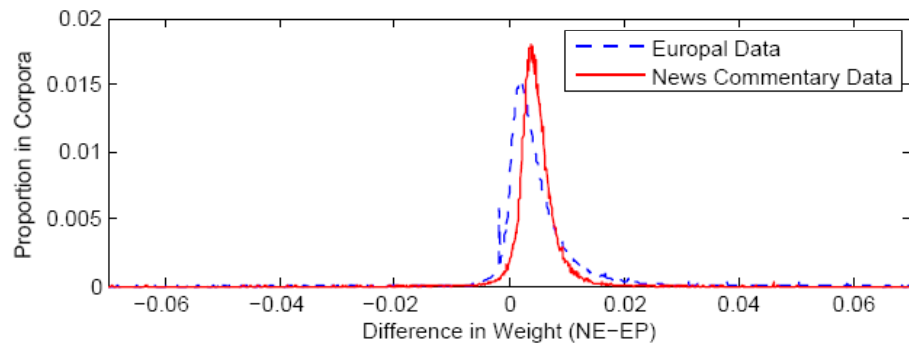
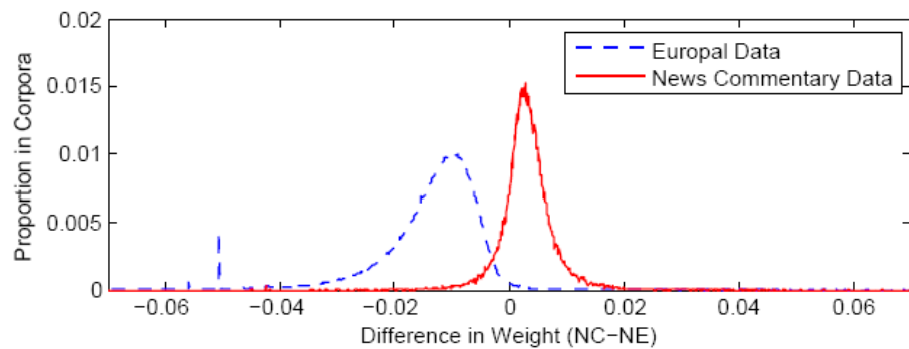
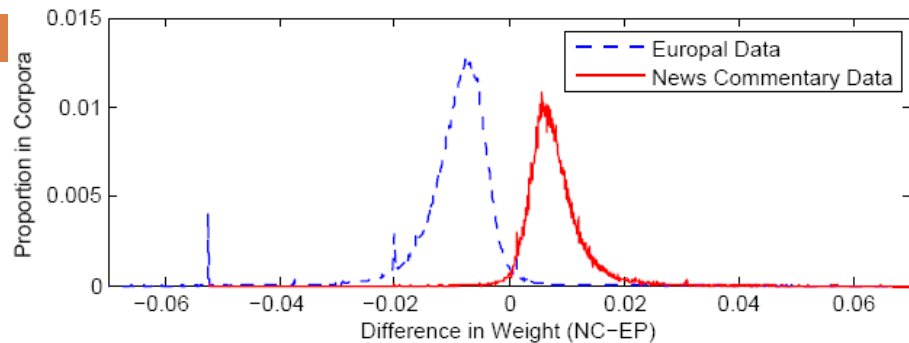
	English	Spanish
Europarl (E)		
sentence pairs	1,258,778	
unique sent. pairs	1,235,134	
avg. sentence length	27.9	29.0
# words	35.14 M	36.54 M
vocabulary	108.7 K	164.8 K
News-Commentary (NC)		
sentence pairs	64,308	
unique sent. pairs	64,205	
avg. sentence length	24.0	27.4
# words	1.54 M	1.76 M
vocabulary	44.2 K	56.9 K

Histogram of sc weights



- Calculated sc for the whole training data using NC, EP and NC+EP(NE) LMs.
- Many sentences get a much higher score in training than using MLE

Histogram of weight differences



- Calculated $gdsc$ for Europal and News-Commentary training data using NC, EP and NC+EP(NE) LMs.
- For each sentence we computed the difference of $gdsc$ between NC and EP LM, namely $gdsc_{NC} - gdsc_{EP}$, and plot histogram.
- Similar analysis have been perform on NC-NE and NE-EP.

IBM Model 4 train perplexities when using Sentence Pair Confidence scores

- IBM Model-4 train perplexities on train and test data

		None	EP+ NC	NC	EP
Train	En \rightarrow Es	46.76	42.36	42.97	44.47
	Es \rightarrow En	70.18	62.81	62.95	65.86
Test	EP (En \rightarrow Es)	91.13	90.89	91.84	90.77
	NC (En \rightarrow Es)	53.04	53.44	51.09	55.94
	EP (Es \rightarrow En)	126.56	125.96	123.23	122.11
	NC (Es \rightarrow En)	81.39	81.28	78.23	80.33

- Perplexities drop significantly in training data of two translation directions.
- In test sets, perplexities also drop in genre which implied a better word alignment model had been learned.

Performance of sentence pair confidence scores (sc , gdsc)

	E06	E07	NCd	NCt1	NCt2
ES → EN					
None	33.26	33.23	36.06	35.56	35.64
NC+EP	33.23	32.29	36.12	35.47	35.97
NC	33.43	33.39	36.14	35.27	35.68
EP	33.36	33.39	36.16	35.63	36.17
EN → ES					
None	33.33	32.25	35.1	34.08	34.43
NC+EP	33.23	32.29	35.12	34.56	34.89
NC	33.3	32.27	34.91	34.07	34.29
EP	33.08	32.29	35.05	34.52	35.03

- The improvements on News-Commentary sets are obvious, especially on held-out evaluation sets NCt and NCt1; using EP obtained the best performance
- No evidence to show that using genre-dependent confidence will provide better result comparing with general confidence.

Performance of sentence-dependent phrase alignment confidence (sdpc)

	E06	E07	NCd	NC†1	NC†2
ES → EN					
None	33.26	33.23	36.06	35.56	35.64
NC+EP +sdpc	33.54	33.39	36.07	35.38	35.85
NC +sdpc	33.17	33.31	35.96	35.74	36.04
EP +sdpc	33.44	32.87	36.22	35.63	36.09
EN → ES					
None	33.33	32.25	35.1	34.08	34.43
NC+EP +sdpc	33.28	32.45	34.82	33.68	33.86
NC +sdpc	33.13	32.47	34.01	34.34	34.98
EP +sdpc	32.97	32.2	34.26	33.99	34.34

- Across development and held-out sets the gains from *sdpc* are inconsistent

Conclusion

- We developed
 - ▣ sentence pair confidence (*sc*)
 - ▣ *genre-dependent* sentence pair confidence (*gdsc*)
 - ▣ *sentence-dependent* phrase alignment confidence (*sdpc*) scores.
- Using source and target language models to estimate scores.
- Experimental results shown that
 - ▣ Better approximation for empirical probability of sentence pairs. Improvements are obtained by using sentence pair confidence scores; using EP LM gain best scores.
 - ▣ No evidence to show that using *gdsc* will provide better result comparing with general confidence.
 - ▣ Test set model perplexities drop by using *gdsc*, but translation results are going against expectation
 - ▣ Did not observe consistent improvements by using *sdpc*



THANK YOU