

Run commands, notes and motivations

Run commands

Below python script will run for about 80 minutes, during which consume 7.5G memory at most.

```
python answer/align.py -p europarl -f de -n 100000 > output.a
head -1000 output.a > upload.a
```

Notes and motivations

The core solution to this language alignment is to use training data to find a most likely from f_i to e_j .

- Core is EM algorithm. Stage expectation collects counts based on already known and training data, then Maximization stage use viterbi alg. to find the most possible pair.
 - Expectation: $Pr(f|e, t) = \prod_{i=1}^I \sum_{j=1}^J t(f_i|e_j)$
 - Maximization: $argmax_t L(t) = argmax_t \sum_s \log Pr(f^{(s)}|e^{(s)}, t)$
- Baseline implements ibm model 1, trains the model to generate parameters $t(f_i|e_j)$ for each word pair, and uses viterbi algorithm to find the most possible word pair.

$$t_k(f_i|e_j) = \sum_{s=1}^N \sum_{(f_i, e_j) \in (f^{(s)}, e^{(s)})} \frac{count(f_i, e_j, f^{(s)}, e^{(s)})}{count(e_j, f^{(s)}, e^{(s)})}$$

$$count(f_i, e_j, f, e) = \frac{t_{k-1}(f_i|e_j)}{\sum_{a_i=1}^J t_{k-1}(f_i|e_{a_i})}$$

$$count(e_j, f, e) = \sum_{i=1}^I count(f_i, e_j, f, e)$$

- First improvement is to train two models French->English and English->French, and then generate optimal parameters from the intersection of two models.
 $opt_{res} = params(e|f) \cap params(f|e)$
- Modify the baseline from 'go over each word in a sentence' to 'go over the word set of this sentence'. The intuition behind this is simple, IBM model 1 only considers word pair, and pays no attention to the position. Thus, set will quicken training descently.
- Diagonal alignment TODO: @SHUO
- IBM model 2
 - Given a French sentence $f = f_1 \dots f_n$ and English sentence $e = e_1 \dots e_m$, we model alignments of the form $a = a_1 \dots a_n$, where each a_i takes a value from 1 to m , denoting the index of the English word to which the i th French word is aligned. Lexical translation parameters t and this a work together to provide a possibility for each word pair.

$$p(f, a|e) = \prod_{i=1}^n p(a_i = j|i, j, m, n) \times p(f_i|e_{a_i})$$

$$p(a_i = j|i, j, m, n) = \frac{1}{Z_{i,m,n}} e^{\lambda h(i,j,m,n)}$$

$$h(i, j, m, n) = -\left|\frac{i}{n} - \frac{j}{m}\right|$$

- Use IBM model 1 output as model 2 input initialization.
- Smoothing TODO: @Shanghao