A: Theory

The algorithm used on this assignment was an enhanced version of the baseline algorithm. The enhancements are listed below:

- Considered both p(f|e) and p(e|f). The output is the intersection of both sets.
- Added NULL words to the end of each f and e sentences.
- The alignments considered are the nearest to the current index *i*. According to the last sentence on *pg*. 837 of the HMM-Based Word Alignment in Statistical Translation paper (http://aclweb.org/anthology/C/C96/C96-2141.pdf), the difference in the position index is smaller than 3 (for Indoeuropean languages).
- Applied probability smoothing for both p(f|e) and p(e|f). Assuming we have a source vocabulary of size v_f and a target vocabulary of size v_e , the formula for calculating probabilities are smoothed as the followings:

$$P(f|e) = \frac{(count_{f,e} + \delta)}{(count_e + \delta * v_f)}$$
(1)

$$P(e|f) = \frac{(count_{e,f} + \delta)}{(count_f + \delta * v_e)}$$
(2)

Here, δ is a very small constant. We chose it to be 0.01

B: Algorithm

The algorithms for the both the word alignment and decoding are shown in the next page:

Algorithm 1 Espectation Maximization Training

Set of sentences F from source language and Set of sentences E from target language **Output:** Probability distributions P(f|e) and P(e|f) for any source word e and target word f 1: add a last word to represent NULL on each f and e sentences. 2: $v_f \leftarrow$ number of distinct source words and $v_e \leftarrow$ number of distinct target words 3: Initialize all counts to 0 4: Initialize $t1_{f|e} \leftarrow \frac{1}{v_f}$ and $t2_{e|f} \leftarrow \frac{1}{v_e}$ for all possible pair (f, e)5: Initialize $\delta \leftarrow 0.01$ 6: while $Epoch \leq MaxEpoch$ do 7: $Epoch \leftarrow Epoch + 1$ for each pair of sentences (f, e) do 8: 9: for each f_i in f do $c \leftarrow 0$ and $z \leftarrow 0$ 10: for each e_i in e do 11: $z \leftarrow z + t1_{f_i|e_j}$ 12: for each e_i in e do 13: $count_{f_i,e_j} \leftarrow count_{f_i,e_j} + \frac{t1_{f_i|e_j}}{z}$ and $count_{e_j} \leftarrow count_{e_j} + \frac{t1_{f_i|e_j}}{z}$ 14: 15: for each e_i in e do $c \leftarrow 0$ and $z \leftarrow 0$ 16: for each f_i in f do 17: 18: $z \leftarrow z + t2_{e_i|f_i}$ for each f_i in f do 19: $count_{e_i,f_j} \leftarrow count_{e_i,f_j} + \frac{t2_{e_i|f_j}}{z}$ and $count_{f_j} \leftarrow count_{f_j} + \frac{t2_{e_i|f_j}}{z}$ 20: for each pair (f, e) in $count_{f, e}$ do 21: $t1_{f|e} \leftarrow \frac{(count_{f,e} + \delta)}{(count_e + \delta * v_f)}$ 22: 23: for each pair (e, f) in $count_{e, f}$ do $t2_{e|f} \leftarrow \frac{(count_{e,f} + \delta)}{(count_f + \delta * v_e)}$ 24: 25: **return** $t1_{f|e}$ and $t2_{e|f}$ for all e and f

Algorithm 2 Decoding

Probability distributions $t1_{f|e}$ and $t2_{e|f}$ for any source word e and target word f **Output:** Set of aligned word pairs (f_i, e_i) 1: $AlignedWords \leftarrow \emptyset$ 2: **for** each pair of sentences (f, e) **do** Ignore possible *NULL* alignments 3: 4: $list_{FE} \leftarrow \emptyset$ and $list_{EF} \leftarrow \emptyset$ for each f_i in f do 5: $best_j \leftarrow argmax_{e_j}t1_{f_i|e_j} // if p(j) = p(best_j) the nearest distance d(f_aj, e_j) is chosen$ 6: add pair (f_i, e_i) to $list_{FE}$ 7: 8: for each e_i in e do $best_j \leftarrow argmax_{f_i}t2_{e_i|f_i}$ // if $p(j) = p(best_j)$ the nearest distance $d(e_aj, f_j)$ is chosen 9: add pair (f_i, e_i) to $list_{EF}$ 10: add $list_{FE} \cup list_{EF}$ to AlignedWords11: 12: **return** AlignedWords