## A: Theory

The algorithm used on this assignment was the implementation of the baseline merged with the ability of finding null word alignments and the posterior probability, accounting both p(f|e) and p(e|f).

## B: Algorithm

The algorithm for the word alignment is given below:

## Algorithm 1 Word Alignment

```
French/German sentences f and English sentences e
Input:
Output: Aligned words
 1: Get count of English words v_e and French/German words v_f
 2: Initialize dictionaries t_1, t_2, count_e, count_f count_{fe} and count_{ef}
 3: repeat
      Make NULL available
 4:
      for each pair of sentences (f, e) do
 5:
         for each f_i in f do
 6:
           Initialize c = 0 and z = 0
 7:
           for each e_i in e do
 8:
              normalize z = 1.0/v_f if first iteration else z + = t_1[(f_i, e_i)]
 9:
           for each e_i in e do
10:
              normalize c = (1.0/v_f)/z if first iteration else c = t_1[(f_i, e_i)]/z
11:
              increment both count_e[e_j] and count_{fe}[(f_i, e_j)] with the value of c
12:
13:
         for each e_i in e do
           Initialize c = 0 and z = 0
14:
           for each f_i in f do
15:
              normalize z = 1.0/v_e if first iteration else z + = t_2[(e_i, f_i)]
16:
           for each f_i in f do
17:
              normalize c = (1.0/v_e)/z if first iteration else c = t_2[(e_i, f_j)]/z
18:
              increment both count_f[f_i] and count_{ef}[(e_i, f_i)] with the value of c
19:
20:
      for each pair (e, f) in count_{fe} do
         t_1[(f,e)] = (count_{fe}[(f,e)] + 0.1)/(count_e[e] + 0.1 * v_f)
21:
      for each pair (e, f) in count_{ef} do
22:
         t_2[(e, f)] = (count_{ef}[(e, f)] + 0.1)/(count_f[f] + 0.1 * v_e)
23:
24: until # of iterations not 5
25: Print the best word alignments according to the instructions of the baseline and ignore the
    alignments defined as NULL
```