Computational Linguistics

Assignment 4 Word alignments

Sangeet Sagar sasa00001@stud.uni-saarland.de

January 15, 2021

1 Introduction

This assignment implements the IBM Model 1, which is used in statistical machine translation (SMT) to train word alignment model. So IBM models in general are generative models, which break up the translation process into smaller steps and achieve better statistics with simpler models. **IBM Model 1** uses only lexical translation. It ignores any position information (order), resulting in translating multisets of words into multisets of words.

2 Description

The script:

- main.py: this is the main script that you should be running.
- ibm_model1.py: IBM Model 1 class file.

The alignment extraction has been performed in two ways. Lets discuss them:

- One-to-one alignment: For every source (English) token, we only take one target token corresponding to the maximum translation probability score.
- One-to-many alignment: We set a threshold score i.e. alpha and for every source token, we only take target tokens whose translation probability score is equal or greater than the threshold. This results in superior results which has been discussed in further section.

I also had a chance to compare results from IMB model 1 with an off-the-shelf aligner MGIZA. Already, having the compiled version for this library, I used it to generate alignments as given in resultsresultsmgiza_out.txt. These were further processed into an index-index format using the script read_mgiza_alignmetns.py, that is accepted by the evaluation script.

3 Requirements

The scripts have been tested on:

Python: 3.8.3
 Numpy: 1.19.2

3. tqdm: 1.6.3. Install: pip install tqdm

4 Project file structure

```
ibm_model1.py
main.py
README.md
read_mgiza_alignmetns.py
results
grid_alignment_one2many.txt
grid_alignment_one2one.txt
ibm1_one2many_alpha0.3.a
ibm1_one2one.a
mgiza.a
mgiza_out.txt
```

5 Usage

• **Help**: for instructions on how to run the script with appropriate arguments. python main.py -help

Implementation of IBM Model 1, which is used in statistical machine translation to train an alignment model.

```
positional arguments:
 eng f
                        path to source (eng) file
 foreign f
                        path to target (foreign) file
 out dir
                        output dir to save the obtained alignments
optional arguments:
 -h, --help
                        show this help message and exit
 -epochs EPOCHS
                        number of training epochs for EM
 -\text{num\_sents} NUM SENTS
                        number of sentences to train from
 -alpha ALPHA
                         threshold score of translation probability
                         for alignment
 -save model SAVE MODEL save trained model
```

- Run IBM Model 1: Given 100K English ↔ French parallel sentences, run IBM model 1 and generate
 one-to-one word alignments
 python main.py jhu-mt-hw/hw2/data/hansards.e jhu-mt-hw/hw2/data/hansards.f results/
- Run IBM model 1 and generate **one-to-many** word alignments.

 python main.py jhu-mt-hw/hw2/data/hansards.e jhu-mt-hw/hw2/data/hansards.f results/ -alpha 0.30

6 Datatset

Trained on 100K parallel English \leftrightarrow French sentences from Hansard French/English dataset.

7 Runtime

- **Total** runtime: 1148.180 s
- Aligner (IBM model 1) runtime: 1142.530 s
- Alignment extraction runtime: 0.704 s

8 Results

• Baseline

```
Precision = 0.243110
Recall = 0.544379
AER = 0.681684
```

- IBM Model 1
 - one-to-one alignment

```
Precision = 0.904762
Recall = 0.491124
AER = 0.350365
```

- one-to-many alignment: alpha 0.30

```
Precision = 0.854103
Recall = 0.677515
AER = 0.235382382
```

• Off-the-shelf aligner: MGIZA, already having the compiled version of MGIZA, I used it to generate alignments and results were:

```
Precision = 0.639601
Recall = 0.742604
AER = 0.326923
```

9 Glimpse of results

While all alignments (*.a files) and alignment-grids (*.txt files) can be found in results, here is a glimpse of an alignment grid (one-to-one alignment):

```
Alignment 5 KEY: ( ) = guessed, * = sure, ? = possible
                                 l je
                                 | ne
                                 | ai
                                 | jamais
     (?)
                                 | rencontré
                                 | une
                                 | seule
             (*)
                                 | prostituée
           ?
                                 | de
                                 | rue
                                 | qui
                                 | voulait
                      ? ? ?
                                 | exercer
        ( )
                      ? ? ?
                                 | un
                        ? ?
                                | tel
                              | métier
                             (*) | .
   е е
           t o h a o e h
   v t
           r o o n
                           е
           e k
   е
                   t
           е е
   r
                   е
           t r
                   d
```