# **Hidden Markov Model and Viterbi Algorithm**

### **Question 1**

### replace infrequent words with \_RARE\_

- implemented in src/replace\_with\_rare.py
- run:

```
$ python replace_with_rare.py
$ python count_freqs.py ../output/ner_train_with_rare.dat > ../output/ner_with_rare.counts
```

- · output file path:
  - updated data file with RARE: output/ner\_train\_with\_rare.dat
  - updated count file with RARE: output/ner\_with\_rare.counts

### compute emission parameters

• implemented in src/baseline.py as compute\_emission(word, ner)

#### Question 2

## baseline NER tagger

- · implemented in src/baseline.py
- · run:

```
$python baseline.py
```

- · output file path: output/dev\_tagged.predict
- evaluation:
  - o run: \$ python eval\_ne\_tagger.py ../input/ner\_dev.key ../output/dev\_tagged.predict
  - · result:

```
Found 14043 NEs. Expected 5931 NEs; Correct: 3117.
                                       F1-Score
        precision
                       recall
Total:
        0.221961
                       0.525544
                                       0.312106
                       0.231230
PER:
        0.435451
                                      0.302061
ORG:
        0.475936
                       0.399103
                                       0.434146
        0.147750
                                       0.252612
                       0.870229
100:
MISC:
        0.491689
                       0.610206
                                       0.544574
```

### **Question 3**

### compute the log probability of trigrams

- Implemented in src/trigram.py
- Test:
  - I made up ten trigrams and stored them in input/trigrams\_test.dat
  - boundary cases below were included
    - o q (y1|\*, \*)
    - o q (y2|\*, y1)
    - q (STOP|y(n-1), yn)
  - run: \$ python trigram.py
  - output path: output/trigrams\_test.dat

### implement viterbi algorithm

- implemented in src/viterbi.py
- output path: output/sen\_tagged.predict

### evaluation

- \$ python eval\_ne\_tagger.py ../input/ner\_dev.key ../output/sen\_tagged.predict
- · result:

```
Found 4704 NEs. Expected 5931 NEs; Correct: 3643.
                                         F1-Score
         precision
                        recall
         0.774447
                        0.614230
Total:
                                         0.685096
PER:
         0.759749
                        0.593580
                                         0.666463
                                         0.536913
0.776056
ORG:
         0.611855
                         0.478326
         0.876458
LOC:
                        0.696292
MISC:
         0.830065
                        0.689468
                                         0.753262
```