COMP6490 Document Analysis IE Assignment

Q2 Effect of the training size on the test set

1/4 Training sentences classifier on Test A

```
processed 46167 tokens with 3076 phrases; found: 2929 phrases; correct: 1917.

accuracy: 95.52%; precision: 65.45%; recall: 62.32%; FB1: 63.85

LOC: precision: 71.11%; recall: 65.62%; FB1: 68.26 886

MISC: precision: 43.52%; recall: 18.01%; FB1: 25.47 108

ORG: precision: 60.42%; recall: 69.12%; FB1: 64.48 1397

PER: precision: 73.61%; recall: 62.46%; FB1: 67.58 538
```

2/4 Training sentences classifier on Test A

```
processed 46167 tokens with 3076 phrases; found: 2915 phrases; correct: 2033.

accuracy: 96.05%; precision: 69.74%; recall: 66.09%; FB1: 67.87

LOC: precision: 72.51%; recall: 71.15%; FB1: 71.82 942

MISC: precision: 51.54%; recall: 25.67%; FB1: 34.27 130

ORG: precision: 67.28%; recall: 69.21%; FB1: 68.23 1256

PER: precision: 74.62%; recall: 69.09%; FB1: 71.74 587
```

3/4 Training sentences classifier on Test A

```
processed 46167 tokens with 3076 phrases; found: 2923 phrases; correct: 2107.

2 accuracy: 96.34%; precision: 72.08%; recall: 68.50%; FB1: 70.25

LOC: precision: 74.02%; recall: 72.71%; FB1: 73.36 943

MISC: precision: 59.06%; recall: 33.72%; FB1: 42.93 149

ORG: precision: 70.64%; recall: 71.91%; FB1: 71.27 1243

PER: precision: 75.34%; recall: 69.87%; FB1: 72.50 588
```

4/4 Training sentences classifier on Test A

```
processed 46167 tokens with 3076 phrases; found: 2958 phrases; correct: 2278.

vaccuracy: 97.06%; precision: 77.01%; recall: 74.06%; FB1: 75.51

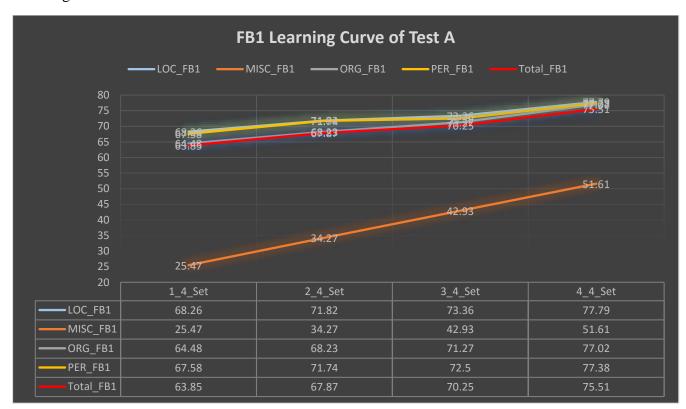
LOC: precision: 80.20%; recall: 75.52%; FB1: 77.79 904

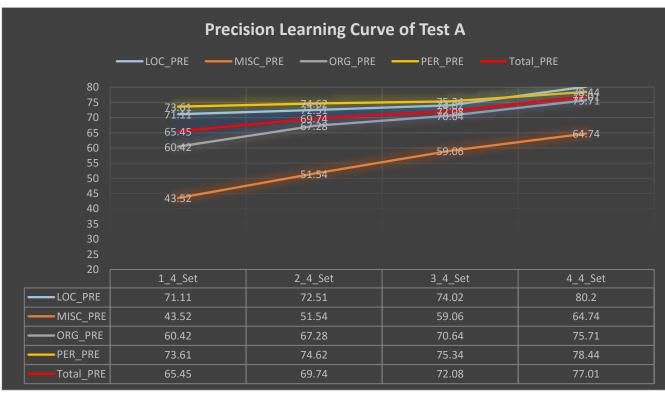
MISC: precision: 64.74%; recall: 42.91%; FB1: 51.61 173

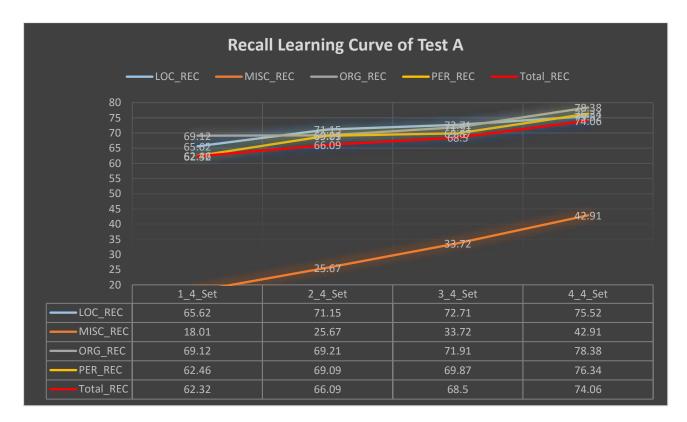
ORG: precision: 75.71%; recall: 78.38%; FB1: 77.02 1264

PER: precision: 78.44%; recall: 76.34%; FB1: 77.38 617
```

Learning Curve of Test A:







1/4 Training sentences classifier on Test B

```
1 processed 47696 tokens with 3877 phrases; found: 3618 phrases; correct: 2170.
2 v accuracy: 93.44%; precision: 59.98%; recall: 55.97%; FB1: 57.91
3 LOC: precision: 53.02%; recall: 65.98%; FB1: 58.79 1094
4 v MISC: precision: 37.14%; recall: 13.68%; FB1: 20.00 140
5 ORG: precision: 60.19%; recall: 62.82%; FB1: 61.48 1575
6 PER: precision: 72.93%; recall: 53.20%; FB1: 61.52 809
7
```

2/4 Training sentences classifier on Test B

```
processed 47696 tokens with 3877 phrases; found: 3605 phrases; correct: 2288.

vaccuracy: 94.12%; precision: 63.47%; recall: 59.01%; FB1: 61.16

LOC: precision: 53.31%; recall: 71.44%; FB1: 61.06 1178

MISC: precision: 44.56%; recall: 22.63%; FB1: 30.02 193

ORG: precision: 67.95%; recall: 60.97%; FB1: 64.27 1354

PER: precision: 74.32%; recall: 58.97%; FB1: 65.76 880
```

3/4 Training sentences classifier on Test B

```
1 processed 47696 tokens with 3877 phrases; found: 3587 phrases; correct: 2360.
2 v accuracy: 94.57%; precision: 65.79%; recall: 60.87%; FB1: 63.24
3 LOC: precision: 54.74%; recall: 74.86%; FB1: 63.24 1202
4 v MISC: precision: 47.85%; recall: 26.32%; FB1: 33.96 209
5 ORG: precision: 70.89%; recall: 61.17%; FB1: 65.67 1302
6 PER: precision: 77.69%; recall: 61.23%; FB1: 68.48 874
```

4/4 Training sentences classifier on Test B

```
processed 47696 tokens with 3877 phrases; found: 3635 phrases; correct: 2660.

LOC: precision: 65.00%; recall: 68.61%; FB1: 70.82

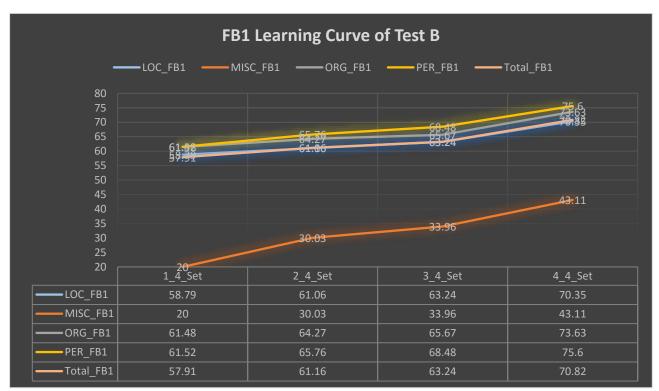
LOC: precision: 65.00%; recall: 76.68%; FB1: 70.35 1037

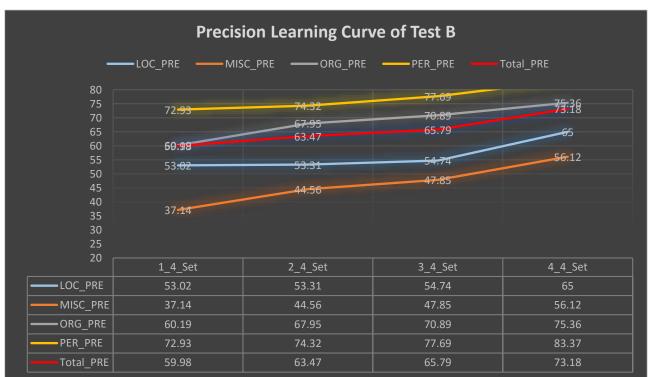
MISC: precision: 56.12%; recall: 35.00%; FB1: 43.11 237

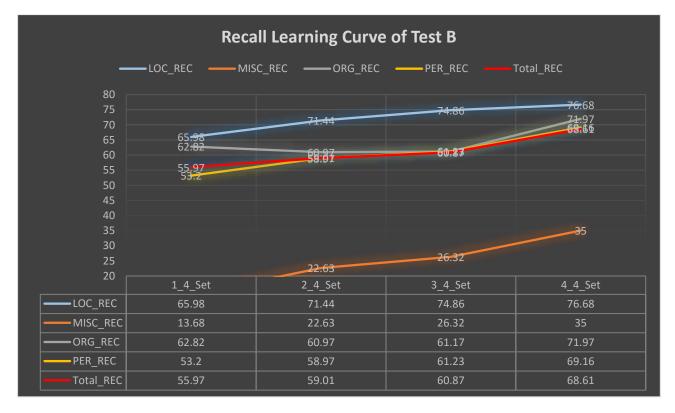
ORG: precision: 75.36%; recall: 71.97%; FB1: 73.63 1441

PER: precision: 83.37%; recall: 69.16%; FB1: 75.60 920
```

Learning Curve of Test B:







Conclusion:

According to the learning curves on test set A and test set B, we can draw the conclusion that the NER performance will improve with the increase of training size.

Q3 Re-substitution performance

Re-substitution is the simplest re-sampling technique to implement. Re-substitution involves using the entire dataset and the test set. [1] The performance of train data using the model trained from itself can be called re-substitution performance.

Re-substitution is usually treated as a poor estimate of generalisation performance because all cases used in the testing have contributed to the data-mining. If the data set is quite representative, resubstitution provides good estimates. The re-substitution method is the simplest re-sampling method to implement and is often used to provide early rough estimates of performance. [1] In this question, I set c value to 0.1, 0.2, 0.3, 0.5, 1.0, 1.5, 2.0, 2.5, 3.0, 10.0. And get the following result of the re-substitution performance on train set.

From the result data and performance curve, we can learn that:

- 1. The re-substitution performance is always pretty high compared with normal performance on other test set.
- 2. The re-substitution performance will improve with the increase of c value. However, the improvement is much more obvious when c value increase from less 0.1 to 1.0. From 1.0, the contribution of increase of c value becomes flat, which because the performance has reached a pretty high state.

Re-substitution Performance on c value 0.1:

```
processed 237201 tokens with 16431 phrases; found: 15785 phrases; correct: 14017.

accuracy: 98.53%; precision: 88.80%; recall: 85.31%; FB1: 87.02

LOC: precision: 87.63%; recall: 86.74%; FB1: 87.18 4388

MISC: precision: 87.21%; recall: 60.87%; FB1: 71.70 1204

ORG: precision: 87.66%; recall: 88.96%; FB1: 88.31 6474

PER: precision: 92.69%; recall: 88.52%; FB1: 90.56 3719
```

Re-substitution Performance on c value 0.2:

```
1 processed 237201 tokens with 16431 phrases; found: 16030 phrases; correct: 15151.
2 v accuracy: 99.27%; precision: 94.52%; recall: 92.21%; FB1: 93.35
3 LOC: precision: 93.77%; recall: 92.98%; FB1: 93.37 4396
4 v MISC: precision: 94.68%; recall: 78.43%; FB1: 85.80 1429
5 ORG: precision: 93.57%; recall: 94.18%; FB1: 93.88 6421
6 PER: precision: 96.93%; recall: 94.20%; FB1: 95.55 3784
```

Re-substitution Performance on c value 0.3:

```
processed 237201 tokens with 16431 phrases; found: 16170 phrases; correct: 15722.

vaccuracy: 99.62%; precision: 97.23%; recall: 95.68%; FB1: 96.45

LOC: precision: 96.32%; recall: 96.28%; FB1: 96.30 4431

MISC: precision: 97.76%; recall: 88.52%; FB1: 92.91 1562

ORG: precision: 96.75%; recall: 96.47%; FB1: 96.61 6361

PER: precision: 98.87%; recall: 96.89%; FB1: 97.87 3816
```

Re-substitution Performance on c value 0.5:

```
processed 237201 tokens with 16431 phrases; found: 16295 phrases; correct: 16079.

accuracy: 99.81%; precision: 98.67%; recall: 97.86%; FB1: 98.26

LOC: precision: 98.43%; recall: 97.81%; FB1: 98.12 4405

MISC: precision: 99.15%; recall: 94.67%; FB1: 96.86 1647

ORG: precision: 98.21%; recall: 98.17%; FB1: 98.19 6376

PER: precision: 99.51%; recall: 98.82%; FB1: 99.16 3867
```

Re-substitution Performance on c value 1.0:

```
processed 237201 tokens with 16431 phrases; found: 16411 phrases; correct: 16292.
vaccuracy: 99.91%; precision: 99.27%; recall: 99.15%; FB1: 99.21
LOC: precision: 99.10%; recall: 99.01%; FB1: 99.05 4429

MISC: precision: 99.65%; recall: 97.74%; FB1: 98.68 1692
ORG: precision: 98.97%; recall: 99.29%; FB1: 99.13 6400
PER: precision: 99.82%; recall: 99.72%; FB1: 99.77 3890
```

Re-substitution Performance on c value 1.5:

```
1 processed 237201 tokens with 16431 phrases; found: 16434 phrases; correct: 16330.
2 accuracy: 99.93%; precision: 99.37%; recall: 99.39%; FB1: 99.38
3 LOC: precision: 99.14%; recall: 99.21%; FB1: 99.18 4436
4 MISC: precision: 99.82%; recall: 98.43%; FB1: 99.12 1701
5 ORG: precision: 99.11%; recall: 99.47%; FB1: 99.29 6402
6 PER: precision: 99.85%; recall: 99.87%; FB1: 99.86 3895
```

Re-substitution Performance on c value 2.0:

```
processed 237201 tokens with 16431 phrases; found: 16437 phrases; correct: 16337.

accuracy: 99.93%; precision: 99.39%; recall: 99.43%; FB1: 99.41

LOC: precision: 99.10%; recall: 99.32%; FB1: 99.21 4443

MISC: precision: 99.82%; recall: 98.49%; FB1: 99.15 1702

ORG: precision: 99.20%; recall: 99.45%; FB1: 99.33 6395

PER: precision: 99.85%; recall: 99.92%; FB1: 99.88 3897
```

Re-substitution Performance on c value 2.5:

```
processed 237201 tokens with 16431 phrases; found: 16432 phrases; correct: 16337.

accuracy: 99.93%; precision: 99.42%; recall: 99.43%; FB1: 99.42

LOC: precision: 99.32%; recall: 99.12%; FB1: 99.22 4424

MISC: precision: 99.65%; recall: 98.72%; FB1: 99.18 1709

ORG: precision: 99.14%; recall: 99.55%; FB1: 99.34 6405

PER: precision: 99.90%; recall: 99.90%; FB1: 99.90 3894
```

Re-substitution Performance on c value 3.0:

```
1 processed 237201 tokens with 16431 phrases; found: 16435 phrases; correct: 16339.
2 v accuracy: 99.93%; precision: 99.42%; recall: 99.44%; FB1: 99.43

LOC: precision: 99.26%; recall: 99.19%; FB1: 99.22 4430

MISC: precision: 99.77%; recall: 98.67%; FB1: 99.21 1706

ORG: precision: 99.17%; recall: 99.51%; FB1: 99.34 6401

PER: precision: 99.85%; recall: 99.95%; FB1: 99.90 3898
```

Re-substitution Performance on c value 10.0:

```
processed 237201 tokens with 16431 phrases; found: 16439 phrases; correct: 16344.

curacy: 99.93%; precision: 99.42%; recall: 99.47%; FB1: 99.45

LOC: precision: 99.14%; recall: 99.35%; FB1: 99.25 4442

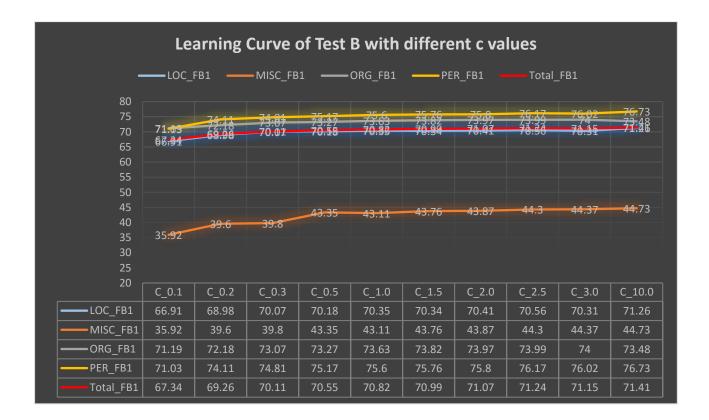
MISC: precision: 99.48%; recall: 99.01%; FB1: 99.24 1717

ORG: precision: 99.31%; recall: 99.42%; FB1: 99.37 6386

PER: precision: 99.90%; recall: 99.90%; FB1: 99.90 3894
```



From the bellowing learn curve, we can see that the improvement on re-substitution performance have some positive effect on general performance. But not obvious enough.



Q4 HMM for POS tagging

1. HMM is a generative probabilistic model, in which a sequence of observable X variables is generated by a sequence of internal hidden states Z. [2]

In this question, the observations can be the single word in the Finnish text, and the hidden states are the six labels: *noun, adjective, pronoun, numeral, particle, verb,* and *other*.

2. The emission probabilities in HMM model refers to the conditional distribution of the observed variables from the specific state. If the observed values Xn are discrete, the probabilities Ø is a K*D table of probabilities, of K hidden states and D symbols (words).

$$B = (b_{ij}) = P(y_i|x_j)$$

Note: x_i is the hidden state, y_i is the observation derived from the hidden state.

In this case, emission probabilities refer to the probabilities of a certain Finnish Symbol derived from a given label out of the six.

3. Transition probability refers to the probability of observing a particular state given that the hidden model is in a particular hidden state.

$$A = (a_{ij}) = P(x_{it}|x_{jt-1})$$

Note: x_i is the hidden state, x_{it-1} is the previous hidden state.

As described in the question, in Finnish, adjectives that define a noun tend to occur before the noun. Which means, in Finnish, the transition probability

$$adj \rightarrow noun >> noun \rightarrow adj$$

$$P(noun|adj) = p_1 \gg P(adj|noun) = p_2$$
.

In Portuguese, on the contrary, the adjective defined a noun tend to occur after the noun.

Then we have:

noun
$$\rightarrow$$
 adj \rightarrow noun
$$P(adj|noun) = p_3 \gg P(noun|adj) = p_4$$

Then we can expect that, the transition probability from an adjective to a noun in Finnish is greater than that in Portuguese, rather than smaller.

References

- [1] A. S. Zeleznikow, "Chapter 9 Evaluation, Deployment," in *Knowledge Discovery from Legal Databases And Related Issues*, Dordrecht, Springer, 2005, p. 174.
- [2] "Tutorial," hmmlearn developers (BSD License), 2016. [Online]. Available: http://hmmlearn.readthedocs.io/en/latest/tutorial.html. [Accessed 13 10 2016].