## 1 Process

- 1. cw-cl-feature:
  - (a) store cw-cl counts into dictionary, delete the frequency that is smaller than 3
  - (b) Phi-1: returns a Python dictionary containing cw-cl counts for each pair of words and tags in the corpus
- 2. pl-cl-feature:
  - (a) same as cw-cl, store pl-cl
- 3. predict
  - (a) take a sentence as input return a prediction of label
  - (b) apply the decimal to 5, to consider all possibility
  - (c) then count the possibility, and output the highest one

## 2 Problem

- 1. accuracy:
  - (a) the accuracy is very low so that I use the multi-pass and shuffle to enhance the performance

## 3 Result

- 1. F1 score:
  - (a) current word\_current label: 0.48
  - (b) previous label\_current label: 0.13
- 2. most positively weighted feature:
  - (a) previous label\_current label:('O\_O', 2959), ('LOC\_O', 940), ('ORG\_O', 752), ('O\_ORG', 99), ('O\_LOC', 78), ('ORG\_ORG', -72), ('PER\_O', -83), ('LOC\_LOC', -115), ('PER\_PER', -140), ('O\_PER', -156)
  - (b) current word\_current label: ('Results\_O', 13), ('-\_O', 11), ('(\_O', 9), ('Men\_O', 9), ('Women\_O', 8), ('-\_O', 8), ('Result\_O', 8), ('Newsroom\_ORG', 8), ('soccer\_O', 8), ('matches\_O', 8)
  - (c) the top list make sense because most are O that are not a namely entity and other ORG looks like an organization as well