

COMS 4705: Natural language Processing  
Programming Assignment 2  
@author: apoorv purwar (UNI - ap3644)

**Code execution instructions –**

- This code has been written in python3.
- For all questions, we will be using parser.py to execute our code.
- For question 4, we used the following commands to generate the required files and generate the result –

- `python parser.py q4 parse_train.dat parse_train.RARE.dat`

- For question 5, we used the following commands to generate the required files and generate the result –

- `python parser.py q5 parse_train.RARE.dat parse_dev.dat q5_prediction_file`

- For question 6, we used the following commands to generate the required files and generate the result –

- `python parser.py q4 parse_train_vert.dat parse_train_vert.RARE.dat`

- `python parser.py q6 parse_train_vert.RARE.dat parse_dev.dat q6_prediction_file`

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**Results and Observations –**

The above commands generated the required files for us and the following results were obtained –

**For Question 4 –**

Type	Total	Precision	Recall	F1 Score
=====				
.	370	1.000	1.000	1.000
ADJ	164	0.827	0.555	0.664
ADJP	29	0.333	0.241	0.280
ADJP+ADJ	22	0.542	0.591	0.565
ADP	204	0.955	0.946	0.951
ADV	64	0.694	0.531	0.602
ADVP	30	0.333	0.133	0.190
ADVP+ADV	53	0.756	0.642	0.694
CONJ	53	1.000	1.000	1.000
DET	167	0.988	0.976	0.982
NOUN	671	0.752	0.842	0.795
NP	884	0.632	0.529	0.576
NP+ADJ	2	0.286	1.000	0.444
NP+DET	21	0.783	0.857	0.818
NP+NOUN	131	0.641	0.573	0.605
NP+NUM	13	0.214	0.231	0.222
NP+PRON	50	0.980	0.980	0.980
NP+QP	11	0.667	0.182	0.286
NUM	93	0.984	0.645	0.779
PP	208	0.588	0.625	0.606
PRON	14	1.000	0.929	0.963
PRT	45	0.957	0.978	0.967
PRT+PRT	2	0.400	1.000	0.571
QP	26	0.647	0.423	0.512
S	587	0.626	0.782	0.695
SBAR	25	0.091	0.040	0.056
VERB	283	0.683	0.799	0.736
VP	399	0.559	0.594	0.576
VP+VERB	15	0.250	0.267	0.258
total	4664	0.714	0.714	0.714

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The execution time for this was – 25.88 seconds

For Question 5 –

Type	Total	Precision	Recall	F1 Score
=====				
.	370	1.000	1.000	1.000
ADJ	164	0.689	0.622	0.654
ADJP	29	0.324	0.414	0.364
ADJP+ADJ	22	0.591	0.591	0.591
ADP	204	0.960	0.951	0.956
ADV	64	0.759	0.641	0.695
ADVP	30	0.417	0.167	0.238
ADVP+ADV	53	0.700	0.660	0.680
CONJ	53	1.000	1.000	1.000
DET	167	0.988	0.994	0.991
NOUN	671	0.795	0.845	0.819
NP	884	0.617	0.548	0.580
NP+ADJ	2	0.333	0.500	0.400
NP+DET	21	0.944	0.810	0.872
NP+NOUN	131	0.610	0.656	0.632
NP+NUM	13	0.375	0.231	0.286
NP+PRON	50	0.980	0.980	0.980
NP+QP	11	0.750	0.273	0.400
NUM	93	0.914	0.688	0.785
PP	208	0.623	0.635	0.629
PRON	14	1.000	0.929	0.963
PRT	45	1.000	0.933	0.966
PRT+PRT	2	0.286	1.000	0.444
QP	26	0.650	0.500	0.565
S	587	0.704	0.814	0.755
SBAR	25	0.667	0.400	0.500
VERB	283	0.790	0.813	0.801
VP	399	0.663	0.677	0.670
VP+VERB	15	0.294	0.333	0.312
total	4664	0.742	0.742	0.742

The execution time in this case was – 49.93 seconds

- Observations with Vertical Markovization (q6) -

Vertical markovization improved our 'total' scores by **0.03** and increased the execution time by approximately **24 seconds**.

The increase in the time was because of the increase in the number of rules, but this increase in the rules made our model robust as we can observe in the increase in the F-Score, Precision and Recall for each of the non-terminals, which can be observed in the tables above for questions 5 and 6.

For both the question, the same code performed efficiently for me as I had used dictionaries in all my variables and wrote the code so that it scales. Hence, there were no changes needed when I ran the same code for Question 6, and it worked efficiently.