Assignment 2: Part of Speech Tagging

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CSE 538 Fall 2018

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Report

Description of Viterbi Implementation:

dp_scores: Viterbi Matrix containing the best score to go (i,j) position in NxL matrix. **best_sequence**: Each best score has different path. This stores the best path to (i,j). It gives the index of the previous tag used for the current best score.

I am iterating from $i \to 1$ to N and $j \to 0$ to L. i is the word & j is the tag being used for the current word. For each (i,j), i am iterating j_each from 0 to L:

Trans_scores[j_each][j] is the cost from j_each tag to j tag. **dp_scores[i-1][j_each]** is the previous best score for the j_each tag having i-1 word.

max_scores = dp_scores[i-1][j_each] + trans_scores[j_each][j]

I store the max_scores in the dp_scores[i][j] best_sequence[i][j] = j_each corresponding to max_scores

After iteration, add the emission score to the current best score at (i,j) position: Update the **dp_score[i][j] += emission_scores[i][j]**

After the iterations are completed, viterbi_index store the position of the best score at the n-1 word after adding end scores to the last column of dp scores.

```
viterbi_index = [np.argmax(dp scores[-1] + end scores)]
```

To get the best sequence, I need to start from the best score in the last column and recursively get the index of the previous tag used for the current best score.

Viterbi_index[-1] is the current last index/tag used. reverse_begin[viterbi_index[-1]] gives the previous best index/tag used.

After getting the list, i need to reverse the list to get the sequence from 1 to N word.

Return (score, list)

Description of Features Added:

I have added the following features beside the basic to feat_gen.py:

- 1.Suffix
- 2.Prefix
- 3. Special Characters (@, ?, !, |, ;), #) & Is the word a Title
- 4. Commonly Used Adjectives, Verb, Preposition, Noun
- 5. X<= 6 implies add "NOT CLASSIFIED" to word having less than or equal to 6 features as we are not able to correctly identify it.
- 1.Suffix: I have labeled some common suffix separately to help it identify noun, adjective, preposition etc.. Like the suffix "ify" will help identify verb. 'ly' & 'wise' suffix to identify adverb. Similary, 'ism' & 'ist' to help it identify Noun.
- 2. Prefix: Similarly to Suffix, i have added some prefix to identify noun, adjective, adverb etc..
- 3. I have added some special characters as they were being frequently used in the twitter text & we need to separate these words being used with special characters as they don't follow the standard grammar rules. I have added @, #, \$, ! etc.. to separately add features corresponding to the character in the word. It will help the classifier to identify it correctly.
- 4.Lastly, i have added some common adjectives, verb, preposition & noun and gave a separate label to it if it contains any of the word. This has tremendously improved the accuracy of the model.
- 5.If the number of features is less than or equal to 6, we identify it as having not classified due to less information about the word & group them together. For eg: "aaaaah" word doesn't make any sense and we can classify it as not classified.

Comparison of New Features vs Basic Features:

<u>Tabel 1 :</u>

Model	Features	Token Wise Accuracy	Precision	Recall	F1 Score
МЕММ	Basic	84.38978	0.85	0.84	0.84
	Basic + Suffix	84.3424	0.85	0.84	0.84
	Suffix+Prefix	84.91012298 9	0.85	0.85	0.85
	Suffix+Prefix +Special Char	85.3358561	0.86	0.85	0.85
	Suffix+Prefix +Special Char + Common Adj.	85.52507	0.86	0.86	0.85
	Suffix+Prefix +Special Char + Common Adj. + Verb	85.95080	0.87	0.86	0.86
	Suffix+Prefix +Special Char + Common Adj. + verb + Preposition	85.95080	0.87	0.86	0.86
	Suffix+Prefix +Special Char + Common Adj.+Verb+P re. + Noun	86.0927	0.87	0.86	0.86
Best	Suffix+Prefi x+Special Char+Adj.+ Verb+Pre. + Noun + X<=6	86.14001892 147589	0.87	0.86	0.86

<u>Tabel 2 :</u>

Model	Features	Token Wise Accuracy	Precision	Recall	F1 Score
CRF	Basic	84.2951	0.84	0.84	0.84
	Basic + Prefix	84.626	0.85	0.85	0.84
	Suffix+Prefix	85.0047	0.85	0.85	0.85
	Suffix+Prefix +Special Char	85.9035	0.86	0.86	0.86
	Suffix+Prefix +Special Char + Common Adj.	86.6130	0.87	0.87	0.87
Best	Suffix+Prefi x+Special Char + Common Adj. + Verb	86.66035	0.87	0.87	0.87
	Suffix+Prefix +Special Char + Common Adj. + verb + Preposition	85.95080	0.86	0.86	0.86
	Suffix+Prefix +Special Char + Common Adj.+Verb+P reposition + Noun	86.56575	0.87	0.87	0.86
	Suffix+Prefix +Special Char+Adj.+V erb+Pre. + Noun + X<=6	86.61305581 835383	0.87	0.87	0.87

For the MEMM Model, we can see that the model with features "Suffix+Prefix+Special Char + Common Adj.+Verb+Preposition + Noun + X<=6" has the best accuracy of 86.0927.

The above model is better than basic MEMM model by having a higher accuracy of 86.140018921 - 84.38978 = **1.750238**

For the CRF model, we can see that the model with features "Suffix+Prefix+Special Char + Common Adj. + Verb" has the best accuracy of 86.56575.

The above model is better than basic MEMM model by having a higher accuracy of 86.66035 - 84.2951 = **2.36525**

From the above table, we observe that CRF model performs better on dev dataset than the MEMM model.

Also, the commonly used **Adjectives + Verb + Noun + Preposition** significantly improves the accuracy of the model.

Note: For more detail reports, please refer to the screenshots at the end of the report.

Difference in Tagging between MEMM with Basic & Best Features:

Sentence: What a productive day. Not . Really hope i can get to **@glasgowfilm** for winter's Bone - need to get on with job applications tonight **then**! Trailer: http://bit.ly/bhUlum

Word	Correct Tag	MEMM with Best Features	Basic Features MEMM
@glasgowfilm	х	Х	VERB
then	ADV	ADV	ADP

Comparison of MEMM & CRF Model:

From the table 1 & table 2, we can see that CRF model performs better than MEMM model with the difference in accuracy being **86.66035 - 86.140018921 = 0.520331079**. We also observer that for the same set of features in the model, the CRF tends to outperform MEMM model in almost all cases with the exception of model having features "

Suffix+Prefix+Special Char + Common Adj. + verb + Preposition". In this case, both the model gives almost same accuracy against the dev dataset.

So, we can safely say that CRF is better than MEMM model.

Difference in Tagging Words CRF Vs MEMM:

Sentence: The dollar held near its highest in a month ... http://bit.ly/a0F3dO

Word	Correct Tag	CRF	МЕММ
The	DET	DET	DET
dollar	NOUN	NOUN	NOUN
held	VERB	NOUN	NOUN
near	ADP	ADP	NOUN
its	PRON	PRON	PRON
highest	ADJ	ADJ	NOUN
in	ADP	ADP	ADP
а	DET	DET	DET
month	NOUN	NOUN	NOUN
http://bit.ly/a0F3dO	х	х	Х

CRF Model:

Basic Feauters:

Token-wise acci	uracy 84.2	951750236	5185	
Token-wise F1	(macro) 83	.21108699	638205	
Token-wise F1	(micro) 84	.29517502	365185	
Sentence-wise	accuracy 1	1.6071428	57142858	
рі	recision	recall	f1-score	support
			120.022	
	0.95	0.98		254
ADJ	0.64	0.55	0.59	99
ADP	0.86	0.87	0.87	151
ADV	0.83	0.62	0.71	129
CONJ	0.95	0.93	0.94	42
DET	0.96	0.91	0.93	130
NOUN	0.79	0.86	0.82	479
NUM	0.85	0.68	0.75	34
PRON	0.99	0.93	0.96	194
PRT	0.84	0.84	0.84	57
VERB	0.79	0.84	0.82	362
X	0.80	0.78	0.79	183
avg / total	0.84	0.84	0.84	2114

Basic Features + Prefix:

```
### Dev evaluation
Token-wise accuracy 84.62630085146643
Token-wise F12 (macro) 83.58081832253595 0. 0.]]
Token-wise F1 (micro) 84.62630085146643
Sentence-wise accuracy 8.928571428571429
             precision
                          recall f1-score support
                  0.95
                            0.98
                                      0.97
                                                 254
        ADJ
                9 0.64
                            0.47
                                      0.54
                                                  99
                                                 151
      ADP
             min 0.84
                           0.87
                                      0.85
                  0.85
                                      0.71
     ADV
                            0.60
                                                 129
            0109740.97
                           0.90
                                                  42
     SCONJ.
                                      0.94
     tivDETearning0:97
                         0.91
                                      0.94
                                                 130
     NOUN
                  0.79
                            0.86
                                      0.82
                                                 479
                            0.74
     OSSNUM.
                0.81
                                      0.77
                                                  34
                            0.95
                                                 194
                0.97
                                      0.96
     PRON
       PRT
                  0.91
                            0.89
                                      0.90
                                                  57
       VERB 0081290.80 1
                           0.83
                                      0.82
                                                 362
         Xearning0.80e:
                            0.81
                                      0.81
                                                 183
avg /ototal.0073160.85[[1.]
                                      0.84
                            0.85
                                                2114
```

Added Suffix+Prefix features:

	ccuracy 85.0 1 (macro) 84			
	1 (micro) 85			
	e accuracy 1			
	precision			support
	0.95			254
ADJ	0.94	0.99	0.96	254
ADJ	0.64	0.54	0.58	199
ADP	0.86	0.88	0.87	151
CADV	0.83	0.67	0.74	129
CONJ	0.98	0.95	0.96	142
NDET	0.98	0.92	0.95	130
NOUN	0.79	0.85	0.82	479
PNUM	0.86	0.74	0.79	34
PRON	0.97	0.94	0.95	194
VPRT	0.85	0.89	0.87	357
VERB	0.82	0.85	0.83	362
X	0.81	0.77	0.79	183
avg / total	0.85	0.85	0.85	2114

Added Suffix+Prefix+Special Characters :

rohit@rohit-galliumos:~/[###:Dev evaluationoaded.			
Token-wise accuracy 85.96	35004730	369	
Token-wise F1 (macro) 84.			
Token-wise F1 (micro) 85.			
Sentence-wise accuracy 15			
1000 featuprecision.			support
2000 features added.			ouppor.
3000 features a0.97.	0.98	0.97	254
4000 ADJtures a0.65.	0.52		99
5000 ADPtures a0.88.	0.89		151
6000 ADVtures a0.84.	0.71		129
7000CONJtures a0.95.	0.95		42
8000 DETtures a0.97.	0.91	0.94	130
9000NOUNtures a0.79.	0.87	0.83	479
10000NUMatures 0.74d.	0.68	0.71	34
1100PRONatures 0.97d.	0.92	0.94	194
12000PRTatures 0.85d.	0.93	0.89	57
1300VERBatures 0.84d.	0.85	0.84	362
14000 fXatures 0.84d.	0.84	0.84	183
avgb/rtotaleights 078684	0.86	0.86	2114

Added Suffix+Prefix+Special Char+Adjective :

### Dev eval			LLU	
	ccuracy 86.6			
	1 (macro) 85			
Token-wise F	1 (micro) 86	.61305581	835383	
Sentence-wis	e accuracy 14	4.2857142	85714285	
	precision	recall	f1-score	support
	0.97	0.98	0.98	254
ADJ	0.69	0.60	0.64	99
ADP	0.87	0.89	0.88	151
ADV	0.84	0.69	0.76	129
CONJ	1.00	0.90	0.95	42
DET	0.98	0.92	0.94	130
NOUN	0.81	0.88	0.84	479
NUM	0.76	0.65	0.70	34
PRON	0.99	0.94	0.97	194
PRT	0.85	0.89	0.87	57
VERB	0.84	0.85	0.85	362
X	0.83	0.85	0.84	183
avg / total	0.87	0.87	0.87	2114

Added Suffix+Prefix+Special Char+Adj.+Verb:

	(macro) 84	603595080 97085164		
Token-wise F1				
Sentence-wise				
	precision		f1-score	support
	0.96	0.99	0.97	254
ADJ	0.72	0.60	0.65	99
ADP	0.85	0.89	0.87	151
ADV	0.84	0.69	0.76	129
CONJ	0.95	0.93	0.94	42
DET	0.97	0.92	0.94	130
NOUN	0.81	0.88	0.84	479
NUM	0.69	0.71	0.70	34
PRON	0.99	0.94	0.97	194
PRT	0.86	0.86	0.86	57
VERB	0.86	0.85		362
X	0.83	0.85	0.84	183

Added Suffix+Prefix+Special Char+Adj.+Verb+Preposition:

Token-wise	accuracy 85.9	508041627	247	
	F1 (macro) 84			
	F1 (micro) 85			
	se accuracy 1			
	precision		f1-score	support
	0.97	0.98	0.98	254
ADJ	0.64	0.57	0.60	99
ADP	0.88	0.88	0.88	151
ADV	0.87	0.65	0.74	129
CONJ	0.97	0.90	0.94	42
DET	0.96	0.92	0.94	130
NOUN	0.79	0.88	0.83	479
NUM	0.71	0.71	0.71	34
PRON	0.98	0.94	0.96	194
PRT	0.86	0.89	0.88	57
VERB	0.85	0.85	0.85	362
X	0.84	0.82	0.83	183
avg / total	0.86	0.86	0.86	2114

Added Suffix+Prefix+Special Char+Adj.+Verb+Pre. + Noun :

	ccuracy 86.5 1 (macro) 85			
	1 (micro) 86			
	e accuracy 1			
	precision		f1-score	support
	0.97	0.98	0.97	254
ADJ	0.67	0.61	0.63	99
ADP	0.88	0.90	0.89	151
ADV	0.86	0.68	0.76	129
CONJ	1.00	0.90	0.95	42
DET	0.97	0.90	0.94	130
NOUN	0.80	0.88	0.84	479
NUM	0.78	0.74	0.76	34
PRON	0.96	0.94	0.95	194
PRT	0.91	0.89	0.90	57
VERB	0.84	0.86	0.85	362
X	0.84	0.83	0.84	183
avg / total	0.87	0.87	0.86	2114

Added Suffix+Prefix+Special Char+Adj.+Verb + Noun :

	ccuracy 86.1			
	1 (macro) 84			
	1 (micro) 86		14/589	
Sentence-wis	e accuracy 1			
	precision	recall	f1-score	support
	0.97	0.98	0.97	254
ADJ	0.66	0.58	0.62	99
ADP	0.84	0.89	0.87	151
ADV	0.83	0.66	0.74	129
CONJ	1.00	0.90	0.95	42
DET	0.95	0.92	0.93	130
NOUN	0.81	0.87	0.84	479
NUM	0.74	0.68	0.71	34
PRON	0.95	0.94	0.95	194
PRT	0.90	0.91	0.90	57
VERB	0.85	0.85		362
X	0.86	0.85	0.85	183
avg / total	0.86	0.86	0.86	2114

Added Suffix+Prefix+Special Char+Adj.+Verb+Pre. + Noun + X<=6:

### Dev evalua Token-wise acc Token-wise F1 Token-wise F1	uracy 86.6 (macro) 85 (micro) 86	.04497645 .61305581	370041 835383	
Sentence-wise a	accuracy 1 recision			support
	0.97	0.99	0.98	254
ADJ	0.63	0.63		99
ADP	0.89	0.87	0.88	151
ADV	0.87	0.70	0.77	129
CONJ	0.97	0.93		42
DET	0.98	0.91		130
NOUN	0.80	0.87	0.84	479
NUM	0.68	0.68		34
PRON	0.98	0.94	0.96	194
PRT	0.89	0.86		57
VERB	0.85	0.86		362
Х	0.85	0.85	0.85	183
avg / total	0.87	0.87	0.87	2114

LR(MEMM):

Basic Features:

### Dev eval Token-wise a		897824030	2744	
	1 (macro) 83			
	1 (micro) 84			
	e accuracy 8			
	precision		f1-score	support
	0.94	0.98	0.96	254
ADJ	0.73	0.36	0.49	99
ADP	0.92	0.88	0.90	151
ADV	0.94	0.59	0.72	129
CONJ	1.00	0.93	0.96	42
DET	0.99	0.92	0.95	130
NOUN	0.73	0.90	0.80	479
NUM	0.85	0.68	0.75	34
PRON	0.99	0.92	0.96	194
PRT	0.89	0.88	0.88	57
VERB	0.80	0.85	0.82	362
X	0.81	0.77	0.79	183
avg / total	0.85	0.84	0.84	2114

Added Suffix + Basic Features:

### Dev evalua Token-wise accı		424787133	3964	
Token-wise F1				
Token-wise F1	(micro) 84	.34247871	333964	
Sentence-wise	accuracy 8	.03571428	5714286	
			f1-score	support
	0.94	0.98	0.96	cod 254
ADJ	0.74	0.37	0.50	99
ADP	0.91	0.87	0.89	151
ADV	0.94	0.58	0.72	129
CONJ	1.00	0.93	0.96	42
DET	0.99	0.92	0.95	130
NOUN	0.73	0.90	0.80	r cod 479
NUM	0.85	0.68	0.75	34
PRON	0.99	0.93	0.96	194
PRT	0.89	0.88	0.88	57
VERB	0.79	0.85	0.82	362
X	0.81	0.78	0.79	183
avg / total	0.85	0.84	0.84	2114

Added Basic + Prefix + Suffix Features:

### Dev evalua	ion				
Token-wise acci	iracy 84.9	101229895	9319		
Token-wise F1	(macro) 83	.88855422	270251		
Token-wise F1	(micro) 84	.91012298	959319		
Sentence-wise a	accuracy 1	0.7142857	14285714		
рі	recision	recall	f1-score	support	
	0.94	0.98	0.96	ur cod 254	
ADJ	0.73	0.37		99	
ADP	0.91	0.89	0.90	151	
ADV	0.89	0.67	0.77	129	
CONJ	1.00	0.93	0.96	42	
DET	0.99	0.92	0.95	130	
NOUN	0.74	0.90	0.81	ur cod 479	
NUM	0.85	0.68	0.75	34	
PRON	0.99	0.93	0.96	194	
PRT	0.89	0.88	0.88	57	
VERB	0.83	0.84	0.83	362	
X	0.80	0.78	0.79	183	
avg / total	0.85	0.85	0.85	2114	

Added Suffix+Prefix+Special Char:

### Dev evalua Token-wise acc Token-wise F1 Token-wise F1 Sentence-wise	curacy 85.33 (macro) 84. (micro) 85.	34230579 33585619	265991 678334	and the second
	precision			support
ADJ	0.95 0.69	0.99 0.38	0.97 0.49	254 99
ADP	0.93	0.89	0.91	151
ADV ADV	0.89	0.68	0.77	129
CONJ	1.00	0.93	0.96	42
DET	0.99	0.92	0.95	130
NOUN	0.73	0.90	0.81	479
NUM	0.85	0.68	0.75	34
PRON	0.99	0.92	0.96	194
PRT	0.89	0.86	0.88	57
VERB	0.81	0.84	0.83	362
X	0.88	0.80	0.84	183
avg / total	0.86	0.85	0.85	2114

Added Suffix+Prefix+Special Char+Adjective:

### Dev evalu	ation			
Token-wise ac		250709555	3453	
Token-wise F1				
Token-wise F1				
Sentence-wise				
	precision		f1-score	support
	0.94	0.99	0.97	254
ADJ	0.77	0.47	0.59	99
ADP	0.92	0.89	0.91	151
ADV	0.88	0.67	0.76	129
CONJ	1.00	0.93	0.96	42
DET	0.99	0.91	0.95	130
NOUN	0.74	0.90	0.81	479
NUM	0.85	0.68	0.75	34
PRON	0.99	0.92	0.96	194
PRT	0.89	0.86	0.88	57
VERB	0.81	0.83	0.82	362
X	0.88	0.80	0.84	183
avg / total	0.86	0.86	0.85	2114

Added Suffix+Prefix+Special Char+Adj.+Verb:

	macro) 85	508041627 . 28147247		
Token-wise F1 (
Sentence-wise a				
			f1-score	support
	0.94	0.99	0.97	254
ADJ	0.80	0.48	0.60	99
ADP	0.93	0.90	0.91	151
ADV	0.88	0.68	0.77	129
CONJ	1.00	0.93	0.96	42
DET	0.99	0.91	0.95	130
NOUN	0.74	0.91	0.82	479
NUM	0.85	0.68	0.75	34
PRON	0.99	0.92	0.96	194
PRT	0.89	0.86	0.88	57
VERB	0.83	0.84	0.84	362
X	0.87	0.80	0.83	183

Added Suffix+Prefix+Special Char+Adj.+Verb+Preposition:

### Dev evalu Token-wise ac Token-wise F1	curacy 85.9			
Token-wise F1	(micro) 85	.95080416	27247	
Sentence-wise	accuracy 1	2.5		
=	precision	recall	f1-score	support
	0.95	0.99	0.97	254
ADJ	0.79	0.48	0.60	99
ADP	0.92	0.88	0.90	151
ADV	0.89	0.68	0.77	129
CONJ	1.00	0.90	0.95	42
DET	0.99	0.92	0.96	130
NOUN	0.74	0.91	0.82	479
NUM	0.85	0.68	0.75	34
PRON	0.99	0.92	0.96	194
PRT	0.89	0.82	0.85	57
VERB	0.84	0.84	0.84	362
X	0.87	0.80	0.83	183
avg / total	0.87	0.86	0.86	2114

Added Suffix+Prefix+Special Char+Adj.+Verb+Pre. + Noun:

### Dev evalua Token-wise ac	curacy 86.0			
Token-wise F1				
Token-wise F1			178808	
Sentence-wise				
	precision	recall	f1-score	support
	0.95	0.99	0.97	254
ADJ	0.77	0.51	0.61	99
ADP	0.92	0.89	0.90	151
ADV	0.89	0.68	0.77	129
CONJ	1.00	0.90	0.95	42
DET	0.99	0.92	0.95	130
NOUN	0.75	0.91	0.82	479
NUM	0.82	0.68	0.74	34
PRON	0.99	0.93	0.96	194
PRT	0.89	0.82	0.85	57
VERB	0.83	0.85	0.84	362
X	0.88	0.80	0.84	183
avg / total	0.87	0.86	0.86	2114

Added Suffix+Prefix+Special Char+Adj.+Verb + Noun:

### Dev evalua Token-wise acc		454115421	0028	
Token-wise F1	(macro) 85	.36652028	83149	
Token-wise F1				
Sentence-wise	accuracy 1	3.3928571	42857142	
ı	orecision	recall	f1-score	support
	0.94	0.99	0.97	254
ADJ	0.78	0.51	0.61	99
ADP	0.92	0.91	0.91	151
ADV	0.88	0.68	0.77	129
CONJ	1.00	0.93	0.96	42
DET	0.99	0.91	0.95	130
NOUN	0.75	0.91	0.82	479
NUM	0.82	0.68	0.74	34
PRON	0.99	0.92	0.96	194
PRT	0.91	0.86	0.88	57
VERB	0.83	0.84	0.83	362
X	0.87	0.80	0.84	183
avg / total	0.87	0.86	0.86	2114

Added Suffix+Prefix+Special Char+Adj.+Verb+Pre. + Noun + X<=6 :

Classes, IZ . MDJ A	IDP AUV	COMO	DE NOUN
### Dev evaluation .			
Token-wise accuracy 86.14	00189214	7589	
Token-wise F1 (macro) 85.	13101496	173871	
Token-wise F1 (micro) 86.	14001892	147589	
Sentence-wise accuracy 12			
5000 featuprecision.		f1-score	support
6000 features added.			
7000 features a0.95.	0.99	0.97	254
8000 ADJtures a0.77.	0.51	0.61	99
9000 ADPtures a0.92.	0.89	0.90	151
10000ADVatures 0.89d.	0.69	0.78	129
1100CONJatures 1.00d.	0.90	0.95	42
12000DETatures 0.99d.	0.92	0.95	130
1300NOUNatures 0.75d.	0.91	0.82	479
14000NUMatures 0.82d.	0.68	0.74	34
379 148PRON 0.99	0.93	0.96	194
Number oPRTeights 078968	0.82	0.85	57
StartinVERBaining 0.83	0.85	0.84	362
iteration X 0.88	0.80	0.84	183
avg loss: 0.405365 w: [[0. 0.]]
avge/ttotalearning0.87e:	1.00.860	0.86	2114
iteration 1			

Test Predictions Generation:

CRF Model Prediction Result:

CRF Model Prediction	i Result:			
### Train evalu	ation			
Token-wise accu		8710201869	66	
Token-wise F1 (
Token-wise F1 (
Sentence-wise a				
		recall f		support
Pi	CCTSTOIL	I CCAIL I	1-30016	Suppor t
	1.00	1.00	1.00	901
ADJ	1.00			341
ADP	1.00			549
ADV	1.00			401
CONJ	1.00			161
DET	1.00			426
NOUN	1.00			The second secon
NUM	0.99			0.74 (0
PRON				B 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
PRON	1.00			10.000000000000000000000000000000000000
VERB			1.00	
VERB X	1.00			The second secon
^	1.00	1.00	1.00	682
avg / total	1 00	1.00	1.00	7381
avy / Local	1.00	1.00	1.00	7301
### Dev evaluat	ion			
Token-wise accu		0350508041	62	
Token-wise F1 (
Token-wise F1 (
Sentence-wise a				
		recall f		support
Ρ'	COISION	I COULT 1	1 30010	Support
	0.96	0.99	0.97	254
ADJ				
ADP	0.85			W1 4 4 5 5 1
ADV	0.84			The state of the s
CONJ	0.95		0.94	7.22
DET	0.97		0.94	
NOUN	0.81	0.88		479
NUM	0.69	0.71	0.70	34
PRON	0.99	0.94	0.97	194
PRT	0.86	0.86	0.86	57
VERB	0.86	0.85	0.86	362
X	0.83	0.85	0.84	183
å				100
avg / total	0.87	0.87	0.87	2114
3				
### Generating	Test predi	ctions		

MEMM Model:

MEMINI MOGCI.				
### Train ev				
Token-wise a	ccuracy 98.5	638802330	3076	
Token-wise F	1 (macro) 98	3.25985972	692378	
Token-wise F	1 (micro) 98	3.56388023	303076	
Sentence-wise accuracy 77.04485488126649				
	precision			support
	1.00	1.00	1.00	901
ADJ			0.96	341
ADP	0.97	0.98	0.98	549
ADV				401
CONJ		0.98		161
DET		1.00		426
NOUN	0.97	0.99	0.98	1685
NUM			0.96	
PRON			1.00	
PRT			0.99	
VERB			0.99	
X	0.99			
avg / total	0.00	0.99	0.99	7381
avy / LULAI	0.99	0.99	0.99	1301
### Dev evaluation Token-wise accuracy 86.14001892147589 Token-wise F1 (macro) 85.13101496173871 Token-wise F1 (micro) 86.14001892147589 Sentence-wise accuracy 12.592857142857142				
Selltellee-MIS	precision			support
	precision	rccarr	11 30010	Suppor C
	0.95	0.99	0.97	254
ADJ			0.61	
ADP		0.89		
ADV		0.69		A CONTRACTOR OF THE CONTRACTOR
CONJ			0.95	
DET			0.95	
NOUN	0.75		0.82	
NUM	0.82		0.74	34
PRON	0.99	0.93	0.96	194
PRT	0.89	0.82	0.85	57
VERB	0.83	0.85	0.84	362
X	0.88	0.80	0.84	183
,	0.00	0.00	0.01	100
avg / total	0.87	0.86	0.86	2114
### Generating Test predictions				