

Report

Writeup:

1. Describe how you handled unknown words in hmm1.

Answer: the words with count less than 3 in the train set was changed to ****UNK**** with its tag also it was checked if the word matches on of pattern. If yes, we add it to the counter of the pattern. In test phase if the word is unknown, we searched if there is a pattern which matches the word if not we took tags from ****UNK**** words

2. Describe your pruning strategy in the Viterbi hmm.

Answer: Each sentence we start only with two tags: "START", "START", after each word we update the first tag by set it to the second tag options and the second tag set we update with the best tag of the current word for Greedy but for Viterbi we take only tags where this word was associated with. We also choosing the best tag from the set if we have the word in dictionary or from pattern set or if there is no pattern we choose all tags

3. Report your test scores when running each tagger (hmm-greedy, hmm-viterbi, maxent-greedy, memm-viterbi) on each dataset. For the NER dataset, report token accuracy accuracy, as well as span precision, recall and F1.

Accuracy on POS:

HMM greedy 0.9405918292648591

HMM viterbi 0.9592589464200242

MEMM greedy 0.9584424360165554

MEMM viterbi 0.9642706309654532

Accuracy on NER:

HMM Greedy

Accuracy: 0.9449183760518051

All-types	Prec:	0.7919586868314276	Rec:	0.722652305621003		
ORG	Prec:	0.6345875542691751	Rec:	0.6539895600298284	F-Score:	0.6441424899008446
LOC	Prec:	0.849197247706422	Rec:	0.8062057702776265	F-Score:	0.8271432560737224
MISC	Prec:	0.7687861271676301	Rec:	0.7212581344902386	F-Score:	0.7442641298265249
PER	Prec:	0.8881900768693222	Rec:	0.6900108577633007	F-Score:	0.7766575007638252

HMM Viterbi

Accuracy: 0.9566481833339796

All-types	Prec:	0.8497678737233055	Rec:	0.7701110737125547		
MISC	Prec:	0.8416075650118203	Rec:	0.7722342733188721	F-Score:	0.8054298642533938
LOC	Prec:	0.9057057057057057	Rec:	0.8209036472509527	F-Score:	0.8612221587664193
ORG	Prec:	0.7156644394951744	Rec:	0.7188665175242357	F-Score:	0.7172619047619049
PER	Prec:	0.9115913555992141	Rec:	0.755700325732899	F-Score:	0.8263579697239538

MEMM Greedy

Accuracy: 0.962425840474621

All-types	Prec:	0.868260144524736	Rec:	0.7886233591383373		
MISC	Prec:	0.867948717948718	Rec:	0.7342733188720173	F-Score:	0.7955346650998825
ORG	Prec:	0.8175618073316283	Rec:	0.715137956748695	F-Score:	0.7629276054097057
PER	Prec:	0.854006586169045	Rec:	0.8447339847991314	F-Score:	0.8493449781659389
LOC	Prec:	0.9210850801479655	Rec:	0.8132825258573761	F-Score:	0.8638334778837815

MEMM Viterbi

Accuracy: 0.9728178680832913

All-types	Prec:	0.8776290630975143	Rec:	0.8497139010434197		
PER	Prec:	0.8914316125598722	Rec:	0.9093376764386536	F-Score:	0.9002956194571351
LOC	Prec:	0.9300578034682081	Rec:	0.8758845944474687	F-Score:	0.9021586767591814
MISC	Prec:	0.879372738238842	Rec:	0.7906724511930586	F-Score:	0.8326670474014849
ORG	Prec:	0.7878326996197719	Rec:	0.7725577926920209	F-Score:	0.7801204819277109

4. Is there a difference in behavior between the hmm and maxent taggers? discuss.

Answer: MEMM is normalized globally and HMM is normalized locally, MEMM using Logistic Regression model which slow down the performance.

5. Is there a difference in behavior between the datasets? discuss.

Answer: Yes, there is a huge number of tags with name "O" and less other tags with location, organization etc, but in POS there is a lot of examples for each tag. In addition, the amount of tags in POS is much bigger than in NER.

6. What will you change in the hmm tagger to improve accuracy on the named entities data?

Answer: We have got a good accuracy on hmm, but we can improve it with more examples of MISC,LOC,PER and ORG or probably add other tags which can be used as probability to NER.

7. What will you change in the memm tagger to improve accuracy on the named entities data, on top of what you already did?

Answer: We change the word pattern and remove patterns which have all options, because it doesn't give us nothing

8. Why are span scores lower than accuracy scores?

Answer: Because number of 'O' general occurrences is much bigger, and we get better accuracy because we right on them instead of span.