

Dataset Characterstics:

1. Tags and Their Frequencies

PER_Others 9087
PER_Victim 1281
PER_Accused 4338
ORG_Victim 400
ORG_Accused 2958
ORG_Others 11715
LOC_Accused 814
LOC_Others 8958
LOC_Event 6806
LOC_Victim 223

Hidden Markov Model Based Named Entity Recognition

Experiment 1: No Feature. 1st order markov order

Hidden states:

Output:

N-gram:

Dealing with Unknowns:

Probability Calculations:

<HiddenMarkovModelTagger 10 states and 20252 output symbols>

['ORG_Others', 'PER_Others', 'LOC_Event', 'PER_Victim', 'LOC_Others',
'PER_Accused', 'ORG_Victim', 'ORG_Accused', 'LOC_Accused', 'LOC_Victim']

For few documents, Number of tags and words are different.

===== Accuracy Class Wise =====

Class	Matched	Total	%
PER_Others:	156	1988	7.85%
PER_Victim:	21	368	5.71%
PER_Accused:	63	743	8.48%
ORG_Victim:	12	96	12.5%
ORG_Accused:	109	537	20.30%
ORG_Others:	246	2340	10.51%
LOC_Accused:	0	145	0.0%
LOC_Others:	198	1762	11.24%
LOC_Event:	154	1111	13.86%
LOC_Victim:	0 29	0.0%	

Average Accuracy: 9.04%

CRF

1. Using NLTK

===== Accuracy Class Wise =====

Class	Matched	Total	%
PER_Others:	1257 1702	73.85%	
PER_Victim:	28 268	10.45%	
PER_Accused:	205 599	34.22%	
ORG_Victim:	0 81	0.0%	
ORG_Accused:	379 546	69.41%	
ORG_Others:	1289 2015	63.97%	
LOC_Accused:	0 120	0.0%	
LOC_Others:	961 1743	55.13%	
LOC_Event:	475 1100	43.18%	
LOC_Victim:	0 27	0.0%	

Average Accuracy: 35.02%

2. Using CRFSuite

===== Accuracy Class Wise =====

Class	Matched	Total	%
PER_Others:	1278 1669	76.57%	
PER_Victim:	58 211	27.48%	
PER_Accused:	291 504	57.74%	
ORG_Victim:	3 45	6.67%	
ORG_Accused:	431 528	81.63%	
ORG_Others:	1277 1957	65.25%	
LOC_Accused:	7 80	8.75%	
LOC_Others:	1000 1625	61.54%	
LOC_Event:	517 1168	44.26%	
LOC_Victim:	2 41	4.88%	

Avg Accuracy = 43.48%

3. Adding Entity Score to CRF with word score calculated using word frequency and position:

$$word_{score} = \log_{10} \left((1 + word_{freq} / word_{total}) * (1 + word_{total} - word_{position}) / word_{total} \right)$$

===== Accuracy Class Wise =====

Class	Matched	Total	%
PER_Others:	1174 1629	72%	
PER_Victim:	64 229	27%	
PER_Accused:	355 884	40%	
ORG_Victim:	3 57	5%	
ORG_Accused:	394 546	72%	
ORG_Others:	1137 2199	51%	
LOC_Accused:	1 160	0%	
LOC_Others:	868 1640	52%	
LOC_Event:	456 1106	41%	
LOC_Victim:	3 31	9%	

Avg Accuracy = 36.2%

LSTM

With word embedding created using Glove from dataset itself.

With pre trained Golve word embedding taken from Stannford NLP.

BLSTM+Softmax

1. With word embedding created using Glove from dataset itself.

===== Accuracy Class Wise =====

Class	Matched	Total	%
PER_Others:	658	823	79%
PER_Victim:	49	182	26%
PER_Accused:	182	428	42%
ORG_Victim:	1	34	2%
ORG_Accused:	234	317	73%
ORG_Others:	655	1145	57%
LOC_Accused:	3	75	4%
LOC_Others:	528	907	58%
LOC_Event:	284	680	41%
LOC_Victim:	0	12	0%

Average Accuracy: 38.2%

2. With pre trained Golve word embedding taken from Stannford NLP.

Class	Matched	Total	%
PER_Others:	664	823	80%
PER_Victim:	50	182	27%
PER_Accused:	232	428	54%
ORG_Victim:	5	34	14%
ORG_Accused:	245	317	77%
ORG_Others:	679	1145	59%
LOC_Accused:	4	75	5%
LOC_Others:	492	907	54%
LOC_Event:	372	680	54%
LOC_Victim:	0	12	0%

Average Accuracy: 42.4%

BLSTM + CRF

1. With word embedding created using Glove from dataset itself.

===== Accuracy Class Wise =====

Class	Matched	Total	%
PER_Others:	665	823	80%
PER_Victim:	28	182	15%
PER_Accused:	161	428	37%
ORG_Victim:	0	34	0%

ORG_Accused: 238 317 75%
ORG_Others: 662 1145 57%
LOC_Accused: 3 75 4%
LOC_Others: 550 907 60%
LOC_Event: 234 680 34%
LOC_Victim: 0 12 0%

Average Accuracy: 36.2%

2. With pre trained Golve word embedding taken from Stannford NLP.

===== Accuracy Class Wise =====

Class	Matched	Total	%
PER_Others:	662	823	80%
PER_Victim:	42	182	23%
PER_Accused:	253	428	59%
ORG_Victim:	3	34	8%
ORG_Accused:	245	317	77%
ORG_Others:	688	1145	60%
LOC_Accused:	1	75	1%
LOC_Others:	489	907	53%
LOC_Event:	371	680	54%
LOC_Victim:	0	12	0%

Average Accuracy: 40%

Analysis of Dataset and Results

1. Context Keyword: Context keyword plays an important role in identifying the tag. In our dataset there is a high overlapping of surrounding words for 'relevant' and 'non-relevant' entities.