

# Nueral Discourse Relation Classification Report

Wei Lu

November 4, 2018

In this project, I am trying to classify the discourse relation between two adjacent or non-adjacent discourse units using neural network models. I am very grateful to the TAs Yuchen Zhang and Che Yeol Chun, who helped me understand the course material more deeply and provided very helpful advice for my implementation. I particularly acknowledge the discussion with Che Yeol who made me realize the importance the connective words, which improved the model performance a lot.

## 1 Feature representation, neural network architecture, and hyper-parameters design

### 1.1 Feature representation

I used the embeddings of words to represent each sentence which is composed of connective (5 words), arg1 (30 words) and arg2 (30 words). So I truncated each sentence into a 65-word sentence with first 5 words as connective (0's are padding when the number of connective words is less than 5). I tried both random embeddings and pretrained google embeddings and found that the random embeddings works a little better. (This is elaborated in the experiments in the third part of the report)

### 1.2 Neural network architecture

I used a one layer fully connected neural network with 300 nodes and used cross entropy with L2 regularizers as the objective function. The optimizer is AdamOptimizer.

### 1.3 Hyper-parameters design

The batch size is 32, learning rate for AdamOptimizer is 0.0001. Regularizers coefficient is 0.01. Word embed size is 300.

## 2 Code structure

I defined the following functions:

```
def vocab_sense(rows):
```

which extracts the vocabulary and all the senses which appeared in the training set. A dictionary was used to store the vocabulary with word as key and its index as value. (When I used pretrained embedding, the corresponding value is the vector of the word).

```
def row2vec(row,vocab):
```

which featurizes a sentence composed of arg1, connective and arg2. It changes each word in the sentence into its corresponding index or 0 if it a word unseen.

```
def main(train_x_list, train_y_list, dev_x_list, dev_y_list, test_x_list,  
test_y_list, vocab_size):
```

which is the main part of the code. It is composed of two parts: first, it builds a feed-forward neural network architecture (symbolic programming). Secondly is the training part, which defines the cross entropy loss, optimizer (AdamOptimizer with L2 regularizers), and does the training.

You can run the code in the terminal with

"python relations\_random\_embedding\_regul.300nodes\_conn5\_arg1.30\_arg2.30.py", then a json file named "output\_random\_embedding\_regul.300nodes\_conn5\_arg1.30\_arg2.30.json" will be generated and can be used in evaluation by running the scorer.py programme.

## 3 Experiments

### 3.1 feature representation

I tried two methods for word embedding: self trained and pretrained google embeddings, and found that the self trained embedding words a little better (although intuitively the pretrained embeddings should be more meaningful). For the length of each sentence I did two experiments: one is truncating each sentence using the first 30 words, another one is paying more attention to the connective words (after I discussed it with Che Yeol). Specifically, each sentence is represented by the first 30 words from arg1 and 5 words from connective, then 30 words from arg2. This is from the information of the training set: the average length of arg1, arg2 and connective is 16.6 15.6 and 1.1. More importantly it aligns all the sentences and shows that word31 to word35 are connective in each sentence. (Another way to do it is to move connective to the beginning of the sentence to make sure that it appears in the representation in case that the truncation is too short). When the connective are made to appear for sure, the performance improves a lot. Another experiment is to select the last 30 words from arg1 instead of the first 30 words since intuitively the last 30 words should play more role in discourse relation, however the results shown are similar. (The experimental results can be found at the end of the report, the last two are the best performance)

### 3.2 neural network architecture

Two architectures have been tried, one is a network with a single hidden layer with 50, 150 or 300 nodes; another one is with two hidden layers with nodes 256(first layer) -128(second layer), or 512(first layer) - 256(second layer). According to the experimental results, for the pretrained embeddings, two layers worded better than one layer and nodes 256 - 128 worked better than 512 - 256 due to the overfitting problem. For the self-trained embedding, the performance of one layer model and two layers model are similar, and 300 nodes worked better than 50 and 150 nodes.

### 3.3 regularizers

The L2 regularizers make the performance improve a lot by punishing the parameters to get rid of the overfitting problem.

### 3.4 experimental results

one layer network with 50 nodes and pretrained embedding without regularizers with sentence length 30:

=====				
Evaluation for all discourse relations				
Accuracy: 0.2531				
Temporal.Asynchronous.Succession	precision 0.33333	recall 0.015385	F1 0.029412	
EntRel	precision 0.11765	recall 0.03	F1 0.047809	
Comparison.Concession	precision 0.0	recall 0.0	F1 -	
Temporal.Synchrony	precision 0.27273	recall 0.054545	F1 0.090909	
Expansion.Conjunction	precision 0.27495	recall 0.88218	F1 0.41924	
Expansion.Restatement	precision 0.0	recall 0.0	F1 -	
Contingency.Cause.Result	precision 0.0	recall 0.0	F1 -	
Contingency.Cause.Reason	precision 0.16667	recall 0.01087	F1 0.020408	
Comparison.Contrast	precision 0.048387	recall 0.051724	F1 0.05	
Expansion.Instantiation	precision 0.0	recall 0.0	F1 -	
Contingency.Condition	precision 0.0	recall 0.0	F1 -	
Expansion.Alternative	precision 0.0	recall 0.0	F1 -	
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -	
Micro-Average	precision 0.2531	recall 0.24113	F1 0.24697	
=====				
Evaluation for explicit discourse relations only				
Accuracy: 0.34892				
Temporal.Asynchronous.Succession	precision 0.33333	recall 0.015385	F1 0.029412	
Comparison.Concession	precision 0.0	recall 0.0	F1 -	
Temporal.Synchrony	precision 0.33333	recall 0.057692	F1 0.098361	
Expansion.Conjunction	precision 0.39451	recall 0.86175	F1 0.54124	
Contingency.Cause.Result	precision 0.0	recall 0.0	F1 -	
Contingency.Cause.Reason	precision 0.33333	recall 0.02	F1 0.037736	
Comparison.Contrast	precision 0.064516	recall 0.068966	F1 0.066667	
Expansion.Restatement	precision 0.0	recall 0.0	F1 -	
Contingency.Condition	precision 0.0	recall 0.0	F1 -	
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -	
Expansion.Instantiation	precision 0.0	recall 0.0	F1 -	
Expansion.Alternative	precision 0.0	recall 0.0	F1 -	
Micro-Average	precision 0.36742	recall 0.31908	F1 0.34155	
=====				
Evaluation for non-explicit discourse relations only				
Accuracy: 0.17152				
EntRel	precision 0.26087	recall 0.03	F1 0.053812	
Expansion.Restatement	precision 0.0	recall 0.0	F1 -	
Contingency.Cause.Reason	precision 0.0	recall 0.0	F1 -	
Expansion.Conjunction	precision 0.17857	recall 0.92105	F1 0.29915	
Contingency.Cause.Result	precision 0.0	recall 0.0	F1 -	

Comparison.Concession	precision 0.0	recall 0.0	F1 -
Comparison.Contrast	precision 0.032258	recall 0.034483	F1 0.033333
Expansion.Instantiation	precision 0.0	recall 0.0	F1 -
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.17152	recall 0.16944	F1 0.17047

two layers network with 256 - 128 nodes and pretrained embedding with L2 regularizers with sentence length 30:

=====			
Evaluation for all discourse relations			
Accuracy: 0.31348			
Temporal.Asynchronous.Succession	precision 0.5	recall 0.2	F1 0.28571
EntRel	precision 0.32472	recall 0.44	F1 0.37367
Comparison.Concession	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.14286	recall 0.018182	F1 0.032258
Expansion.Conjunction	precision 0.35839	recall 0.61934	F1 0.45404
Expansion.Restatement	precision 0.32143	recall 0.11842	F1 0.17308
Contingency.Cause.Result	precision 0.090909	recall 0.029412	F1 0.044444
Contingency.Cause.Reason	precision 0.23611	recall 0.18478	F1 0.20732
Comparison.Contrast	precision 0.061224	recall 0.10345	F1 0.076923
Expansion.Instantiation	precision 0.45455	recall 0.11364	F1 0.18182
Contingency.Condition	precision 0.42308	recall 0.40741	F1 0.41509
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.27083	recall 0.26	F1 0.26531
Micro-Average	precision 0.31348	recall 0.29866	F1 0.30589
=====			
Evaluation for explicit discourse relations only			
Accuracy: 0.36331			
Temporal.Asynchronous.Succession	precision 0.8125	recall 0.2	F1 0.32099
Comparison.Concession	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.16667	recall 0.019231	F1 0.034483
Expansion.Conjunction	precision 0.51079	recall 0.65438	F1 0.57374
Contingency.Cause.Result	precision 0.0	recall 0.0	F1 -
Contingency.Cause.Reason	precision 0.30435	recall 0.28	F1 0.29167
Comparison.Contrast	precision 0.071429	recall 0.13793	F1 0.094118
Expansion.Restatement	precision 0.11765	recall 0.33333	F1 0.17391
Contingency.Condition	precision 0.5	recall 0.40741	F1 0.44898
Temporal.Asynchronous.Precedence	precision 0.39394	recall 0.325	F1 0.35616
Expansion.Instantiation	precision 1.0	recall 0.66667	F1 0.8
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.41822	recall 0.33224	F1 0.3703
=====			

# Evaluation for non-explicit discourse relations only

Accuracy: 0.27106

EntRel	precision 0.44444	recall 0.44	F1 0.44221
Expansion.Restatement	precision 0.41026	recall 0.10959	F1 0.17297
Contingency.Cause.Reason	precision 0.11538	recall 0.071429	F1 0.088235
Expansion.Conjunction	precision 0.21429	recall 0.55263	F1 0.30882
Contingency.Cause.Result	precision 0.13333	recall 0.05	F1 0.072727
Comparison.Concession	precision 0.0	recall 0.0	F1 -
Comparison.Contrast	precision 0.047619	recall 0.068966	F1 0.056338
Expansion.Instantiation	precision 0.33333	recall 0.073171	F1 0.12
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.277	recall 0.26778	F1 0.27231

two layers network with 256 - 128 nodes and pretrained embedding with L2 regularizers with 30words\_arg1  
5words\_connective 30words\_arg2:

## ===== Evaluation for all discourse relations

Accuracy: 0.47725

Temporal.Asynchronous.Succession	precision 0.97917	recall 0.72308	F1 0.83186
EntRel	precision 0.33179	recall 0.715	F1 0.45325
Comparison.Concession	precision 0.84615	recall 0.098214	F1 0.176
Temporal.Synchrony	precision 0.58491	recall 0.56364	F1 0.57407
Expansion.Conjunction	precision 0.49559	recall 0.67976	F1 0.57325
Expansion.Restatement	precision 0.6	recall 0.039474	F1 0.074074
Contingency.Cause.Result	precision 0.71429	recall 0.14706	F1 0.2439
Contingency.Cause.Reason	precision 0.68293	recall 0.30435	F1 0.42105
Comparison.Contrast	precision 0.18571	recall 0.22414	F1 0.20312
Expansion.Instantiation	precision 0.75	recall 0.068182	F1 0.125
Contingency.Condition	precision 0.88889	recall 0.88889	F1 0.88889
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.85714	recall 0.72	F1 0.78261
Micro-Average	precision 0.47804	recall 0.45469	F1 0.46607

## ===== Evaluation for explicit discourse relations only

Accuracy: 0.69065

Temporal.Asynchronous.Succession	precision 0.97917	recall 0.72308	F1 0.83186
Comparison.Concession	precision 0.84615	recall 0.1358	F1 0.23404
Temporal.Synchrony	precision 0.59615	recall 0.59615	F1 0.59615
Expansion.Conjunction	precision 0.86957	recall 0.82949	F1 0.84906
Contingency.Cause.Result	precision 0.75	recall 0.32143	F1 0.45
Contingency.Cause.Reason	precision 0.68293	recall 0.56	F1 0.61538
Comparison.Contrast	precision 0.19118	recall 0.44828	F1 0.26804

Expansion.Restatement	precision 0.5	recall 0.5	F1 0.5
Contingency.Condition	precision 0.88889	recall 0.88889	F1 0.88889
Temporal.Asynchronous.Precedence	precision 0.85714	recall 0.9	F1 0.87805
Expansion.Instantiation	precision 1.0	recall 0.66667	F1 0.8
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.74131	recall 0.63158	F1 0.68206
=====			
Evaluation for non-explicit discourse relations only			
Accuracy: 0.29556			
EntRel	precision 0.36203	recall 0.715	F1 0.48067
Expansion.Restatement	precision 0.75	recall 0.020548	F1 0.04
Contingency.Cause.Reason	precision 0.0	recall 0.0	F1 -
Expansion.Conjunction	precision 0.18219	recall 0.39474	F1 0.24931
Contingency.Cause.Result	precision 0.5	recall 0.025	F1 0.047619
Comparison.Concession	precision 0.0	recall 0.0	F1 -
Comparison.Contrast	precision 0.0	recall 0.0	F1 -
Expansion.Instantiation	precision 0.5	recall 0.02439	F1 0.046512
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.29556	recall 0.29198	F1 0.29376

two layers network with 256 - 128 nodes and pretrained embedding with L2 regularizers with 5words\_connective last 30words\_arg1 30words\_arg2:

=====			
Evaluation for all discourse relations			
Accuracy: 0.47808			
Temporal.Asynchronous.Succession	precision 1.0	recall 0.73846	F1 0.84956
EntRel	precision 0.32697	recall 0.685	F1 0.44265
Comparison.Concession	precision 0.86667	recall 0.11607	F1 0.20472
Temporal.Synchrony	precision 0.58621	recall 0.61818	F1 0.60177
Expansion.Conjunction	precision 0.49241	recall 0.6858	F1 0.57323
Expansion.Restatement	precision 0.4	recall 0.039474	F1 0.071856
Contingency.Cause.Result	precision 0.71429	recall 0.14706	F1 0.2439
Contingency.Cause.Reason	precision 0.77778	recall 0.30435	F1 0.4375
Comparison.Contrast	precision 0.17391	recall 0.2069	F1 0.18898
Expansion.Instantiation	precision 0.75	recall 0.068182	F1 0.125
Contingency.Condition	precision 0.88889	recall 0.88889	F1 0.88889
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.87805	recall 0.72	F1 0.79121
Micro-Average	precision 0.47887	recall 0.45548	F1 0.46688
=====			
Evaluation for explicit discourse relations only			

Accuracy: 0.70144				
Temporal.Asynchronous.Succession	precision 1.0	recall 0.73846	F1 0.84956	
Comparison.Concession	precision 0.86667	recall 0.16049	F1 0.27083	
Temporal.Synchrony	precision 0.58621	recall 0.65385	F1 0.61818	
Expansion.Conjunction	precision 0.87864	recall 0.8341	F1 0.85579	
Contingency.Cause.Result	precision 0.75	recall 0.32143	F1 0.45	
Contingency.Cause.Reason	precision 0.8	recall 0.56	F1 0.65882	
Comparison.Contrast	precision 0.17647	recall 0.41379	F1 0.24742	
Expansion.Restatement	precision 0.42857	recall 0.5	F1 0.46154	
Contingency.Condition	precision 0.88889	recall 0.88889	F1 0.88889	
Temporal.Asynchronous.Precedence	precision 0.87805	recall 0.9	F1 0.88889	
Expansion.Instantiation	precision 1.0	recall 0.66667	F1 0.8	
Expansion.Alternative	precision 0.0	recall 0.0	F1 -	
Micro-Average	precision 0.75145	recall 0.64145	F1 0.6921	
=====				
Evaluation for non-explicit discourse relations only				
Accuracy: 0.2879				
EntRel	precision 0.35677	recall 0.685	F1 0.46918	
Expansion.Restatement	precision 0.375	recall 0.020548	F1 0.038961	
Contingency.Cause.Reason	precision 0.0	recall 0.0	F1 -	
Expansion.Conjunction	precision 0.18039	recall 0.40351	F1 0.24932	
Contingency.Cause.Result	precision 0.5	recall 0.025	F1 0.047619	
Comparison.Concession	precision 0.0	recall 0.0	F1 -	
Comparison.Contrast	precision 0.0	recall 0.0	F1 -	
Expansion.Instantiation	precision 0.5	recall 0.02439	F1 0.046512	
Expansion.Alternative	precision 0.0	recall 0.0	F1 -	
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -	
Temporal.Synchrony	precision 0.0	recall 0.0	F1 -	
Micro-Average	precision 0.2879	recall 0.28442	F1 0.28615	

two layers network with 256 - 128 nodes and self trained embedding with L2 regularizers with 30words\_arg1  
5words\_connective 30words\_arg2:

=====				
Evaluation for all discourse relations				
Accuracy: 0.52605				
Temporal.Asynchronous.Succession	precision 0.89286	recall 0.76923	F1 0.82645	
EntRel	precision 0.31898	recall 1.0	F1 0.48368	
Comparison.Concession	precision 1.0	recall 0.071429	F1 0.13333	
Temporal.Synchrony	precision 0.52727	recall 0.52727	F1 0.52727	
Expansion.Conjunction	precision 0.86905	recall 0.66163	F1 0.75129	
Expansion.Restatement	precision 0.57143	recall 0.026316	F1 0.050314	
Contingency.Cause.Result	precision 0.92857	recall 0.19118	F1 0.31707	
Contingency.Cause.Reason	precision 0.80645	recall 0.27174	F1 0.4065	
Comparison.Contrast	precision 0.21333	recall 0.27586	F1 0.2406	
Expansion.Instantiation	precision 0.83333	recall 0.11364	F1 0.2	

Contingency.Condition	precision 0.88889	recall 0.88889	F1 0.88889
Expansion.Alternative	precision 1.0	recall 0.33333	F1 0.5
Temporal.Asynchronous.Precedence	precision 0.86364	recall 0.76	F1 0.80851
Micro-Average	precision 0.52693	recall 0.50118	F1 0.51373
=====			
Evaluation for explicit discourse relations only			
Accuracy: 0.75899			
Temporal.Asynchronous.Succession	precision 0.89286	recall 0.76923	F1 0.82645
Comparison.Concession	precision 1.0	recall 0.098765	F1 0.17978
Temporal.Synchrony	precision 0.52727	recall 0.55769	F1 0.54206
Expansion.Conjunction	precision 0.90213	recall 0.97696	F1 0.93805
Contingency.Cause.Result	precision 0.90909	recall 0.35714	F1 0.51282
Contingency.Cause.Reason	precision 0.80645	recall 0.5	F1 0.61728
Comparison.Contrast	precision 0.21333	recall 0.55172	F1 0.30769
Expansion.Restatement	precision 0.6	recall 0.5	F1 0.54545
Contingency.Condition	precision 0.88889	recall 0.88889	F1 0.88889
Temporal.Asynchronous.Precedence	precision 0.88372	recall 0.95	F1 0.91566
Expansion.Instantiation	precision 1.0	recall 0.66667	F1 0.8
Expansion.Alternative	precision 1.0	recall 0.5	F1 0.66667
Micro-Average	precision 0.76311	recall 0.69408	F1 0.72696
=====			
Evaluation for non-explicit discourse relations only			
Accuracy: 0.32772			
EntRel	precision 0.31949	recall 1.0	F1 0.48426
Expansion.Restatement	precision 0.5	recall 0.0068493	F1 0.013514
Contingency.Cause.Reason	precision 0.0	recall 0.0	F1 -
Expansion.Conjunction	precision 0.41176	recall 0.061404	F1 0.10687
Contingency.Cause.Result	precision 1.0	recall 0.075	F1 0.13953
Comparison.Concession	precision 0.0	recall 0.0	F1 -
Comparison.Contrast	precision 0.0	recall 0.0	F1 -
Expansion.Instantiation	precision 0.75	recall 0.073171	F1 0.13333
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.32772	recall 0.32375	F1 0.32572

one layer network with 50 nodes and self trained embedding with L2 regularizers with 5words\_connective 30words\_arg1 30words\_arg2:

=====			
Evaluation for all discourse relations			
Accuracy: 0.5244			
Temporal.Asynchronous.Succession	precision 0.97917	recall 0.72308	F1 0.83186



EntRel	precision 0.31496	recall 1.0	F1 0.47904
Comparison.Concession	precision 0.73333	recall 0.098214	F1 0.17323
Temporal.Synchrony	precision 0.5625	recall 0.49091	F1 0.52427
Expansion.Conjunction	precision 0.90213	recall 0.64048	F1 0.74912
Expansion.Restatement	precision 0.66667	recall 0.039474	F1 0.074534
Contingency.Cause.Result	precision 0.83333	recall 0.22059	F1 0.34884
Contingency.Cause.Reason	precision 0.58333	recall 0.30435	F1 0.4
Comparison.Contrast	precision 0.2027	recall 0.25862	F1 0.22727
Expansion.Instantiation	precision 1.0	recall 0.090909	F1 0.16667
Contingency.Condition	precision 0.96296	recall 0.96296	F1 0.96296
Expansion.Alternative	precision 1.0	recall 0.33333	F1 0.5
Temporal.Asynchronous.Precedence	precision 0.95	recall 0.76	F1 0.84444
Micro-Average	precision 0.5257	recall 0.49961	F1 0.51232
=====			
Evaluation for explicit discourse relations only			
Accuracy: 0.76259			
Temporal.Asynchronous.Succession	precision 0.97917	recall 0.72308	F1 0.83186
Comparison.Concession	precision 0.78571	recall 0.1358	F1 0.23158
Temporal.Synchrony	precision 0.5625	recall 0.51923	F1 0.54
Expansion.Conjunction	precision 0.90909	recall 0.96774	F1 0.9375
Contingency.Cause.Result	precision 0.91667	recall 0.39286	F1 0.55
Contingency.Cause.Reason	precision 0.59574	recall 0.56	F1 0.57732
Comparison.Contrast	precision 0.2027	recall 0.51724	F1 0.29126
Expansion.Restatement	precision 0.8	recall 0.66667	F1 0.72727
Contingency.Condition	precision 0.96296	recall 0.96296	F1 0.96296
Temporal.Asynchronous.Precedence	precision 0.97436	recall 0.95	F1 0.96203
Expansion.Instantiation	precision 1.0	recall 0.66667	F1 0.8
Expansion.Alternative	precision 1.0	recall 0.5	F1 0.66667
Micro-Average	precision 0.76812	recall 0.69737	F1 0.73103
=====			
Evaluation for non-explicit discourse relations only			
Accuracy: 0.32159			
EntRel	precision 0.31546	recall 1.0	F1 0.47962
Expansion.Restatement	precision 0.5	recall 0.013699	F1 0.026667
Contingency.Cause.Reason	precision 0.0	recall 0.0	F1 -
Expansion.Conjunction	precision 0.5	recall 0.017544	F1 0.033898
Contingency.Cause.Result	precision 0.66667	recall 0.1	F1 0.17391
Comparison.Concession	precision 0.0	recall 0.0	F1 -
Comparison.Contrast	precision 0.0	recall 0.0	F1 -
Expansion.Instantiation	precision 1.0	recall 0.04878	F1 0.093023
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.32159	recall 0.3177	F1 0.31963

one layer network with 150 nodes and self trained embedding with L2 regularizers with 5words\_connective

30words\_arg1 30words\_arg2:

=====				
Evaluation for all discourse relations				
Accuracy: 0.52936				
Temporal.Asynchronous.Succession	precision 1.0	recall 0.70769	F1 0.82883	
EntRel	precision 0.31348	recall 1.0	F1 0.47733	
Comparison.Concession	precision 0.9375	recall 0.13393	F1 0.23437	
Temporal.Synchrony	precision 0.55172	recall 0.58182	F1 0.56637	
Expansion.Conjunction	precision 0.89873	recall 0.6435	F1 0.75	
Expansion.Restatement	precision 0.63636	recall 0.046053	F1 0.08589	
Contingency.Cause.Result	precision 0.8125	recall 0.19118	F1 0.30952	
Contingency.Cause.Reason	precision 0.66667	recall 0.26087	F1 0.375	
Comparison.Contrast	precision 0.24324	recall 0.31034	F1 0.27273	
Expansion.Instantiation	precision 1.0	recall 0.068182	F1 0.12766	
Contingency.Condition	precision 0.92857	recall 0.96296	F1 0.94545	
Expansion.Alternative	precision 1.0	recall 0.33333	F1 0.5	
Temporal.Asynchronous.Precedence	precision 0.97436	recall 0.76	F1 0.85393	
Micro-Average	precision 0.53024	recall 0.50433	F1 0.51696	
=====				
Evaluation for explicit discourse relations only				
Accuracy: 0.77878				
Temporal.Asynchronous.Succession	precision 1.0	recall 0.70769	F1 0.82883	
Comparison.Concession	precision 0.9375	recall 0.18519	F1 0.30928	
Temporal.Synchrony	precision 0.5614	recall 0.61538	F1 0.58716	
Expansion.Conjunction	precision 0.90987	recall 0.97696	F1 0.94222	
Contingency.Cause.Result	precision 0.91667	recall 0.39286	F1 0.55	
Contingency.Cause.Reason	precision 0.66667	recall 0.48	F1 0.55814	
Comparison.Contrast	precision 0.24658	recall 0.62069	F1 0.35294	
Expansion.Restatement	precision 0.8	recall 0.66667	F1 0.72727	
Contingency.Condition	precision 0.92857	recall 0.96296	F1 0.94545	
Temporal.Asynchronous.Precedence	precision 0.97436	recall 0.95	F1 0.96203	
Expansion.Instantiation	precision 1.0	recall 0.66667	F1 0.8	
Expansion.Alternative	precision 1.0	recall 0.5	F1 0.66667	
Micro-Average	precision 0.78442	recall 0.71217	F1 0.74655	
=====				
Evaluation for non-explicit discourse relations only				
Accuracy: 0.317				
EntRel	precision 0.31447	recall 1.0	F1 0.47847	
Expansion.Restatement	precision 0.5	recall 0.020548	F1 0.039474	
Contingency.Cause.Reason	precision 0.0	recall 0.0	F1 -	
Expansion.Conjunction	precision 0.25	recall 0.0087719	F1 0.016949	
Contingency.Cause.Result	precision 0.5	recall 0.05	F1 0.090909	
Comparison.Concession	precision 0.0	recall 0.0	F1 -	
Comparison.Contrast	precision 0.0	recall 0.0	F1 -	

Expansion.Instantiation	precision 1.0	recall 0.02439	F1 0.047619
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.317	recall 0.31316	F1 0.31507

one layer network with 300 nodes and self trained embedding with L2 regularizers with 5words\_connective  
30words\_arg1 30words\_arg2: (one of the best performance)

=====			
Evaluation for all discourse relations			
Accuracy: 0.53184			
Temporal.Asynchronous.Succession	precision 1.0	recall 0.70769	F1 0.82883
EntRel	precision 0.31397	recall 1.0	F1 0.4779
Comparison.Concession	precision 0.92308	recall 0.10714	F1 0.192
Temporal.Synchrony	precision 0.6	recall 0.65455	F1 0.62609
Expansion.Conjunction	precision 0.89873	recall 0.6435	F1 0.75
Expansion.Restatement	precision 0.6	recall 0.039474	F1 0.074074
Contingency.Cause.Result	precision 0.875	recall 0.20588	F1 0.33333
Contingency.Cause.Reason	precision 0.65789	recall 0.27174	F1 0.38462
Comparison.Contrast	precision 0.24658	recall 0.31034	F1 0.27481
Expansion.Instantiation	precision 1.0	recall 0.090909	F1 0.16667
Contingency.Condition	precision 0.92857	recall 0.96296	F1 0.94545
Expansion.Alternative	precision 1.0	recall 0.33333	F1 0.5
Temporal.Asynchronous.Precedence	precision 0.97436	recall 0.76	F1 0.85393
Micro-Average	precision 0.53317	recall 0.5067	F1 0.5196
=====			
Evaluation for explicit discourse relations only			
Accuracy: 0.78058			
Temporal.Asynchronous.Succession	precision 1.0	recall 0.70769	F1 0.82883
Comparison.Concession	precision 0.92308	recall 0.14815	F1 0.25532
Temporal.Synchrony	precision 0.6	recall 0.69231	F1 0.64286
Expansion.Conjunction	precision 0.90987	recall 0.97696	F1 0.94222
Contingency.Cause.Result	precision 0.91667	recall 0.39286	F1 0.55
Contingency.Cause.Reason	precision 0.67568	recall 0.5	F1 0.57471
Comparison.Contrast	precision 0.25	recall 0.62069	F1 0.35644
Expansion.Restatement	precision 0.75	recall 0.5	F1 0.6
Contingency.Condition	precision 0.92857	recall 0.96296	F1 0.94545
Temporal.Asynchronous.Precedence	precision 0.97436	recall 0.95	F1 0.96203
Expansion.Instantiation	precision 1.0	recall 0.66667	F1 0.8
Expansion.Alternative	precision 1.0	recall 0.5	F1 0.66667
Micro-Average	precision 0.78766	recall 0.71382	F1 0.74892
=====			
Evaluation for non-explicit discourse relations only			

Accuracy: 0.32006			
EntRel	precision 0.31496	recall 1.0	F1 0.47904
Expansion.Restatement	precision 0.5	recall 0.020548	F1 0.039474
Contingency.Cause.Reason	precision 0.0	recall 0.0	F1 -
Expansion.Conjunction	precision 0.25	recall 0.0087719	F1 0.016949
Contingency.Cause.Result	precision 0.75	recall 0.075	F1 0.13636
Comparison.Concession	precision 0.0	recall 0.0	F1 -
Comparison.Contrast	precision 0.0	recall 0.0	F1 -
Expansion.Instantiation	precision 1.0	recall 0.04878	F1 0.093023
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.32006	recall 0.31619	F1 0.31811

one layer network with 300 nodes and self trained embedding with L2 regularizers with 5words\_connective  
last 30words\_arg1 30words\_arg2: (one of the best performance)

=====			
Evaluation for all discourse relations			
Accuracy: 0.53267			
Temporal.Asynchronous.Succession	precision 1.0	recall 0.72308	F1 0.83929
EntRel	precision 0.31299	recall 1.0	F1 0.47676
Comparison.Concession	precision 0.82143	recall 0.20536	F1 0.32857
Temporal.Synchrony	precision 0.57143	recall 0.50909	F1 0.53846
Expansion.Conjunction	precision 0.90987	recall 0.64048	F1 0.75177
Expansion.Restatement	precision 0.54545	recall 0.039474	F1 0.07362
Contingency.Cause.Result	precision 0.77778	recall 0.20588	F1 0.32558
Contingency.Cause.Reason	precision 0.58696	recall 0.29348	F1 0.3913
Comparison.Contrast	precision 0.23729	recall 0.24138	F1 0.23932
Expansion.Instantiation	precision 1.0	recall 0.090909	F1 0.16667
Contingency.Condition	precision 0.92857	recall 0.96296	F1 0.94545
Expansion.Alternative	precision 1.0	recall 0.33333	F1 0.5
Temporal.Asynchronous.Precedence	precision 0.97436	recall 0.76	F1 0.85393
Micro-Average	precision 0.534	recall 0.50749	F1 0.5204
=====			
Evaluation for explicit discourse relations only			
Accuracy: 0.78597			
Temporal.Asynchronous.Succession	precision 1.0	recall 0.72308	F1 0.83929
Comparison.Concession	precision 0.82143	recall 0.28395	F1 0.42202
Temporal.Synchrony	precision 0.58333	recall 0.53846	F1 0.56
Expansion.Conjunction	precision 0.90987	recall 0.97696	F1 0.94222
Contingency.Cause.Result	precision 0.91667	recall 0.39286	F1 0.55
Contingency.Cause.Reason	precision 0.6	recall 0.54	F1 0.56842
Comparison.Contrast	precision 0.23729	recall 0.48276	F1 0.31818
Expansion.Restatement	precision 0.8	recall 0.66667	F1 0.72727
Contingency.Condition	precision 0.92857	recall 0.96296	F1 0.94545

Temporal.Asynchronous.Precedence	precision 0.97436	recall 0.95	F1 0.96203
Expansion.Instantiation	precision 1.0	recall 0.66667	F1 0.8
Expansion.Alternative	precision 1.0	recall 0.5	F1 0.66667
Micro-Average	precision 0.7931	recall 0.71875	F1 0.7541
=====			
Evaluation for non-explicit discourse relations only			
Accuracy: 0.317			
EntRel	precision 0.31397	recall 1.0	F1 0.4779
Expansion.Restatement	precision 0.33333	recall 0.013699	F1 0.026316
Contingency.Cause.Reason	precision 0.0	recall 0.0	F1 -
Expansion.Conjunction	precision 0.0	recall 0.0	F1 -
Contingency.Cause.Result	precision 0.5	recall 0.075	F1 0.13043
Comparison.Concession	precision 0.0	recall 0.0	F1 -
Comparison.Contrast	precision 0.0	recall 0.0	F1 -
Expansion.Instantiation	precision 1.0	recall 0.04878	F1 0.093023
Expansion.Alternative	precision 0.0	recall 0.0	F1 -
Temporal.Asynchronous.Precedence	precision 0.0	recall 0.0	F1 -
Temporal.Synchrony	precision 0.0	recall 0.0	F1 -
Micro-Average	precision 0.317	recall 0.31316	F1 0.31507