Supplementary data for

# Document-level biomedical relation extraction via hierarchical tree graph and relation segmentation module

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Availability and implementation: Our code and supplementary data are available at <a href="https://github.com/passengeryjy/HTGRS">https://github.com/passengeryjy/HTGRS</a>.

### **Discussion**

## Impact of hierarchical concept in document graph.

To evaluate the hierarchical concept introduced compared to the traditional document graph, we conduct a series of ablation studies by deleting the different level nodes on the CDR and BioRED dev set. The results are shown in Table S1. Specifically, when we remove the document node from the hierarchical tree graph, we observe that the model's performance drops to varying degrees on both the CDR and BioRED datasets, with drops of 1.2% and 1.1%, respectively. When we delete the sentence nodes and make the document node connect to the mention nodes, it can be seen that the performance only have a slight drop. It is worth noting that the mention information is the most direct source of entity concept features in the document, so we do not remove it from the hierarchical tree graph.

Finally, we pruned the introduced local context nodes, and the model's F1 scores dropped by 0.8% and 1.0% on the CDR and BioRED datasets, respectively. The above results show that the influence of document, local context, and sentence nodes gradually weakens the model performance. We attribute this to the fact that document node representation, as a global feature, contributes significantly to model's ability to relation reasoning based on entity level. Since mentions and local context information are fine-grained models as crucial elements, redundant sentence information is filtered out to a certain extent. In conclusion, our findings support the notion that introducing the concept of hierarchy in document graphs is a valuable approach for modeling entity-level relation reasoning.

Table S1. The impact of node information on CDR and BioRED dev sets.

Model	CDR			BioRED		
	P(%)	R(%)	F1(%)	P(%)	R(%)	F1(%)
HTGRS	84.4	89.4	86.9	59.3	76.8	66.9
w/o doc_node	84.2	87.3	85.7	60.1	72.7	65.8
w/o sen_node	84.2	88.9	86.5	58.2	77.0	66.3
w/o con_node	81.3	89.3	86.1	62.9	69.3	65.9

# Effectiveness analysis of three stage decoupling.

To further verify the effectiveness of three-stage decoupling, we assess the model HTGRS with RSM or without RSM on different groups of dev set in CDR and BioRED, respectively, which are classified into four groups based on the number of entities. Specifically, we use two steps, 5 and 7, to group CDR and BioRED based on the characteristics of the distribution of entities in the document in the data set, respectively. Table S2 represents the group status. As shown in Figure S1, the model's performance with RSM or without RSM sharp drops with increasing entity numbers. As the number of entities in the document increases, the number of entity pairs involved increases rapidly, accompanied by a sharp increase in the complexity of entity pair interaction. We observe that the model with RSM consistently outperforms the model without RSM. When the number of entities increases, the performance of the model without RSM sharply declines, but the model with RSM is the opposite, declining steadily. This result demonstrates that attention to the interactive reasoning between entity pairs is useful for relation prediction, and our relation segmentation module is helpful for the model to conduct interaction reasoning.

Table S2. Proportions of samples in four groups to all dev set.

Datasets	Four groups					
	[1,5)/[1,7)	[5,10)/[7,14)	[10,15)/[14,21)	[15,-)/ [21,-)		
CDR	24.8%	60.6%	12.2%	2.4%		
BioRED	15%	49%	21%	15%		

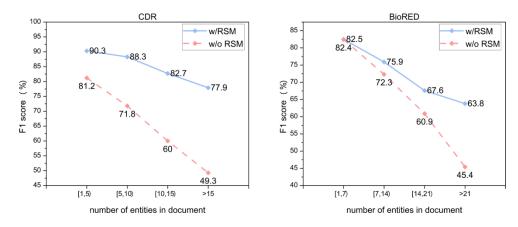


Figure S1. The F1 score in terms of number of entities on CDR and BioRED dev set.

### Case study

In addition, we randomly chose some prediction samples on the CDR dataset to analyze our model HTGRS further. Specifically, we select two of the most striking examples, represented in Table S3. Case 1 shows that for the chemical entity D002110, its mention of caffeine only appears in sentence 1, while the disease entity D007674 appears in sentence 11. This chemical disease entity pair has a positive relationship, but its semantic dependence spans the entire document. So, it is tough for the model to judge the relation correctly. However, HTGRS can identify its relationship as positive, which we attribute to HTGRS fully understanding the global interaction information through entity pair-based relation reasoning. Furthermore, we confirmed this situation by re-predicting the results by removing the RS module, named w/o RSM predictive label; we observe that the predictive result is wrong without RSM.

However, in case 2, since the mention of chemical entity *D002220* and disease entity *D006331* co-occur in the first sentence, the model easily predicts their CID relation as a positive through semantic understanding. However, according to the BioCreative-V community annotation specification, relation instances without specific CID relation tasks do not have CID relation, so the chemical disease entity is negatively related to the real relation. This result shows that our model has certain limitations in scenarios that require background knowledge.

Table S3. Case study of the prediction results on CDR dataset. a

#### Case 1:

PMID: 25986755

Document: [1]Low functional programming of renal AT2R mediates the developmental origin of glomerulosclerosis in adult offspring induced by prenatal caffeine (Chemical:D002110) exposure . [2]UNASSIGNED: Our previous study has indicated that prenatal caffeine (Chemical:D002110) exposure (PCE) could induce intrauterine growth retardation (IUGR) of offspring. ..... [11]These results demonstrated that PCE could induce dysplasia of fetal kidneys (Disease: D007674) as well as glomerulosclerosis of adult offspring, and the low functional programming of renal AT2R might mediate the developmental origin of adult glomerulosclerosis.

Target entity pair: <caffeine, dysplasia of fetal kidneys>

True label: positive relationship

w/ RSM Predictive label: positive relationship √

w/o RSM Predictive label: negative relationship ×

#### Case 2:

**PMID:** 1728915

**Document:** [1]Carbamazepine (Chemical:D002220) - induced cardiac dysfunction (Disease:D006331). [2]Characterization of two distinct clinical syndromes . [3]A patient with sinus bradycardia and atrioventricular block . . . [4]From the analysis of these cases, two distinct forms of carbamazepine - associated cardiac dysfunction (Disease:D006331) emerged . . . . [7]Because carbamazepine . . .

Target entity pair: < Carbamazepine, cardiac dysfunction>

True label: negative relationship

w/ RSM Predictive label: positive relationship ×

w/o RSM Predictive label: positive relationship ×

<sup>&</sup>lt;sup>a</sup>Yellow highlight refers to the mention of head entity as it appears in the document, and blue highlight signifies to the mention of tail entity.