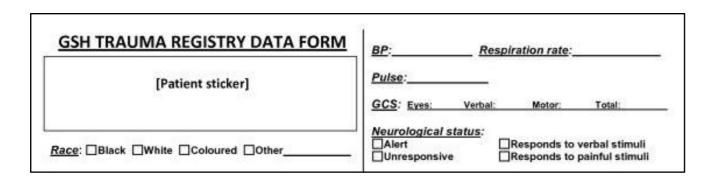


# **Enhancing Visually-Rich Document Understanding via Layout Structure Modeling**

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#### **Limiations of Previus Work**

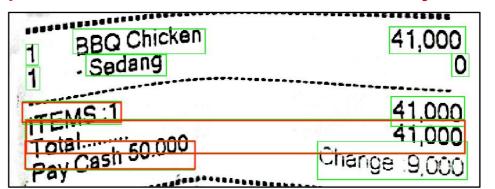
1. Mismatch between the raw order and proper understanding order.



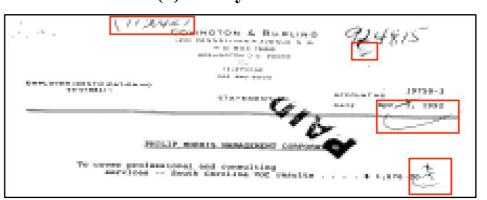
(a) Raw Order: GSH TRAUMA REGISTERY DATA FORM BP: Respiration rate: [Patient sticker] Pulse: GCS: Eyes: Verbal: Motor: Total: Neurological status: Race: Black White Coloured Other Alert Responds to verbal stimuli Unresponsive Responds to painful stimuli

**(b) Proper Order:** GSH TRAUMA REGISTERY DATA FORM [Patient sticker] Race: Black White Coloured Other BP: Respiration rate: Pulse: GCS: Eyes: Verbal: Motor: Total: Neurological status: Alert Responds to verbal stimuli Unresponsive Responds to painful stimuli

2. Inadequate disclosure of document structure by visual information.



(a) Faulty Text Box



(b) Interference Information in Resized Image

## **GraphLayoutLM: Layout Language Model Enhanced by Layout Struture Graph**

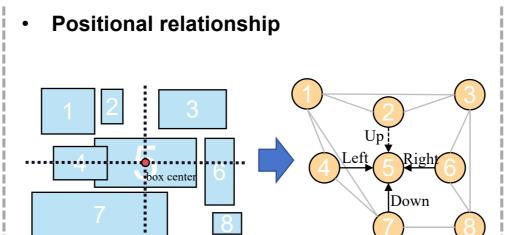
- 1. Layout graph modeling.
- Hierarchical relationship

Parent node identification rule of region *P*:

$$L(n_i) = \begin{cases} 1, & box_i = Top(P) \text{ and } box_i = Left(P) \\ 0, & otherwise \end{cases}$$

Layout Hierarchical subtree construction of region *P*:

$$T_p = \langle n_p, C_p n_p, Parent - Child \rangle$$



Layout structure graph construction

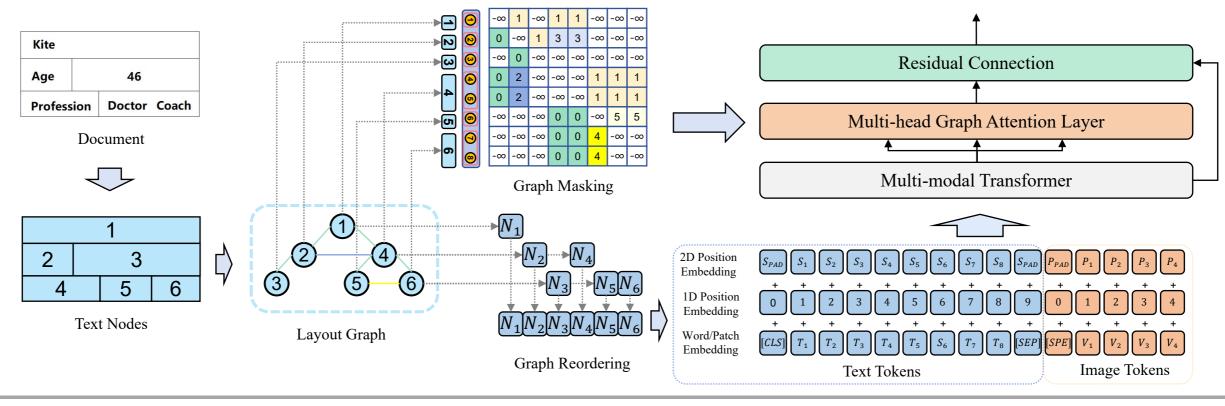
**Hierarchical relationships** are used to build the layout structure tree  $T_g$ .

**Positional relationships** are used to enrich the tree into a layout structure graph *G*.

Representation of *G*:

 $G = T_g \cup \{ < n_i, n_j, rel(n_i, n_j) > | n_i, n_j \in T_g. sibling \}$ 

- 2. The architecture of GraphLayoutLM.
  - GraphLayoutLM utilizes document layout structures graph through Graph Reordering and Graph Masking strategies.



#### **Experiment Results**

#### 1. Experiment results of SER task on English datasets

Model	Parameters	Modality	FUNSD			CORD		
			Precision	Recall	<b>F</b> 1	Precision	Recall	<b>F1</b>
BERT <sub>BASE</sub>	110M	T	54.69	67.10	60.26	88.33	91.07	89.68
RoBERTa <sub>BASE</sub>	125M	T	63.49	69.75	66.48	-	-	93.54
BROS <sub>BASE</sub>	110M	T+L	81.16	85.02	83.05	95.58	95.14	95.3
LayoutLM <sub>BASE</sub>	160M	T+L+I	76.77	81.95	79.27	94.37	95.08	94.7
XYLayoutLM <sub>BASE</sub>	-1	T+L+I	-	-	83.35	-	-	-
LayoutLMv2 <sub>BASE</sub>	200M	T+L+I	80.29	85.37	82.76	94.53	95.39	94.9
DocFormer <sub>BASE</sub>	183M	T+L+I	80.76	86.09	83.34	96.52	96.14	96.3
ERNIE-Layout <sub>BASE</sub>	-1	T+L+I	-	-	90.28	-	-	96.6
LayoutLMv3 <sub>BASE</sub>	133M	T+L+I	-	-	90.29	-	-	96.5
LayoutLMv3 <sub>BASE</sub> †	133M	T+L+I	90.82	91.55	91.19	96.35	96.71	96.5
GraphLayoutLM <sub>BASE</sub> (Ours)	135M	T+L+I+G	92.46	93.85	93.15	97.02	97.53	97.2
BERT <sub>LARGE</sub>	340M	T	61.13	70.85	65.63	88.86	91.68	90.2
RoBERTa <sub>LARGE</sub>	355M	T	67.80	73.91	70.72	-	-	93.8
LayoutLM <sub>LARGE</sub>	343M	T+L	75.96	82.19	78.95	94.32	95.54	94.9
BROS <sub>LARGE</sub>	340M	T+L	82.81	86.31	84.52	_	-	97.2
$StructuralLM_{LARGE}$	355M	T+L	83.52	86.81	85.14	-	-	-
LayoutLMv2 <sub>LARGE</sub>	426M	T+L+I	83.24	85.19	84.20	95.65	96.37	96.0
DocFormer <sub>LARGE</sub>	536M	T+L+I	82.29	86.94	84.55	97.25	96.74	96.9
ERNIE-Layout <sub>LARGE</sub>	-1	T+L+I	-	-	93.12	-	-	97.2
LayoutLMv3 <sub>LARGE</sub>	368M	T+L+I	-	-	92.08	-	-	97.4
LayoutLMv3 <sub>LARGE</sub> <sup>†</sup>	368M	T+L+I	91.51	92.70	92.10	97.45	97.52	97.4
GraphLayoutLM <sub>LARGE</sub> (Ours)	372M	T+L+I+G	94.49	94.30	94.39	97.75	97.75	97.7

### 2. Experiment results of SER task on Chinese datasets

Model	Modality	XFUND F1
XLM-RoBERTa <sub>BASE</sub>	T	87.74
XLM-RoBERTa <sub>LARGE</sub>	T	89.25
LayoutXLM <sub>BASE</sub>	T+L+I	89.24
LayoutXLM <sub>LARGE</sub>	T+L+I	91.61
XYLayoutLM <sub>BASE</sub>	T+L+I	91.76
ERNIE-LayoutX <sub>BASE</sub> <sup>‡</sup>	T+L+I	88.58
LayoutLMv3-Chinese <sub>BASE</sub> <sup>‡</sup>	T+L+I	92.02
LayoutLMv3-Chinese <sub>BASE</sub> †	T+L+I	91.82
GraphLayoutLM-Chinese <sub>BASE</sub> (Ours)	T+L+I+G	93.56

#### 3. Ablation Study

Dataset	Graph Reorder	Graph Mask	Accuracy	Precision	Recall	F1
FUNSD	×	Х	84.76	90.82	91.55	91.19
	1	X	85.70	92.36	93.15	92.75
	×	1	86.75	91.73	92.05	91.89
	1	1	88.39	92.46	93.85	93.15
CORD	×	Х	97.11	96.35	96.71	96.53
	✓	X	97.33	96.79	97.16	96.97
	X	✓	97.88	96.94	97.23	97.09
	✓	✓	98.01	97.02	97.53	97.28
XFUND	×	×	85.87	89.79	93.94	91.82
	✓	X	85.61	89.98	94.37	92.12
	X	✓	90.88	91.58	94.43	92.99
	✓	✓	91.19	91.80	95.38	93.56