### 1. Tabular Inference Problem

- Inference task where premises are tabular.
- Given a premise table determine hypothesis is true (entailment), false (contradiction), or undetermined 2 Add noise and confusion caused by lengthy (neutral), i.e. tabular natural language inference. additions. At times definitions are incorrect.

James Hetfield			
Birth Name	James Alan Hetfield		
Born	Aug. 3, 1963(age 58), California, U.S.		
Genres	Heavy metal, thrash metal, hard rock		
Occupation(s)	Musician, Singer		
Instruments	Vocals, Guitar		
Years active	1978-present		
Labels	Warner Bros, Elektra, MegaForce		
Hypothesis	James Hetfield was born on the west coast of the USA.		

• Example InfoTabS dataset (Gupta et al., 2020), here the Hypothesis is **Entailed**.

### 2. Knowledge Addition

• External knowledge is essential for model reasoning.

Hypothesis	James Hetfield was born on the we coast of the USA.				
Focused Relation	$coast \xleftarrow{AtLocation} california$				
Human	Entailment				
RoBERTa	Neutral				
Trans-KBLSTM	Entailment				

Through a novel architecture, Trans-KBLSTM, this work tackle challenges inherent in prior methods of Knowledge Extraction, Addition, and Integration.

We did a case study on **INFOTABS**, a Tabular NLI Dataset.

2 External knowledge graphs supplement model.

### **3** Challenges and Motivation

- Prior work on knowledge addition for tabular reasoning focus on explicit addition.
- **Knowledge Extraction:** Extract contextually relevant knowledge from external source.
- **Sknowledge Representation:** Effectively represent external semantic knowledge relations.
- **4 Knowledge Integration:** Schematically integrate external knowledge into model design?

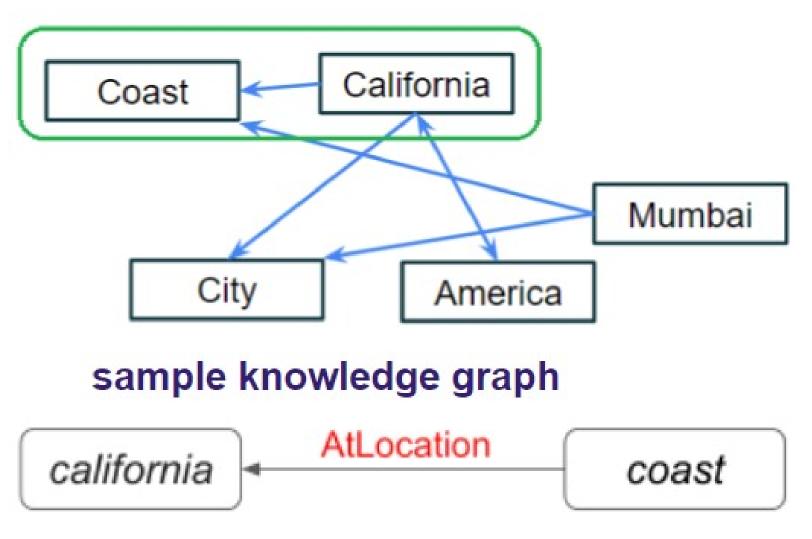
# **Trans-KBLSTM**

Yerram Varun<sup>1\*</sup>, Aayush Sharma<sup>1\*</sup>, Vivek Gupta<sup>2\*</sup>

<sup>1</sup>Indian Institute of Technology, Guwahati; <sup>2</sup>University of Utah

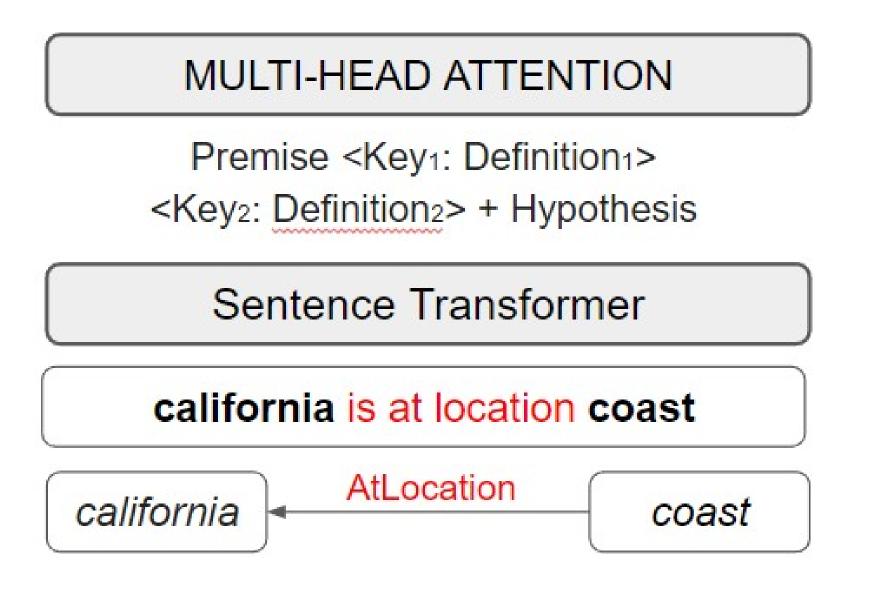
# 4. Knowledge Extraction

• Neeraja et al. 2021 augments the input with lengthy key definitions.



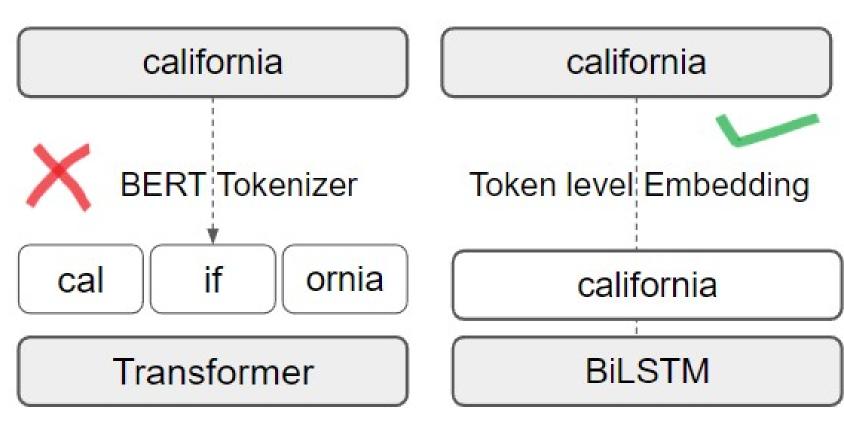
# 5. Knowledge Addition

• Knowledge Triplets are converted to sentences. 2 Sentences are encoded using Sentence Transformers.

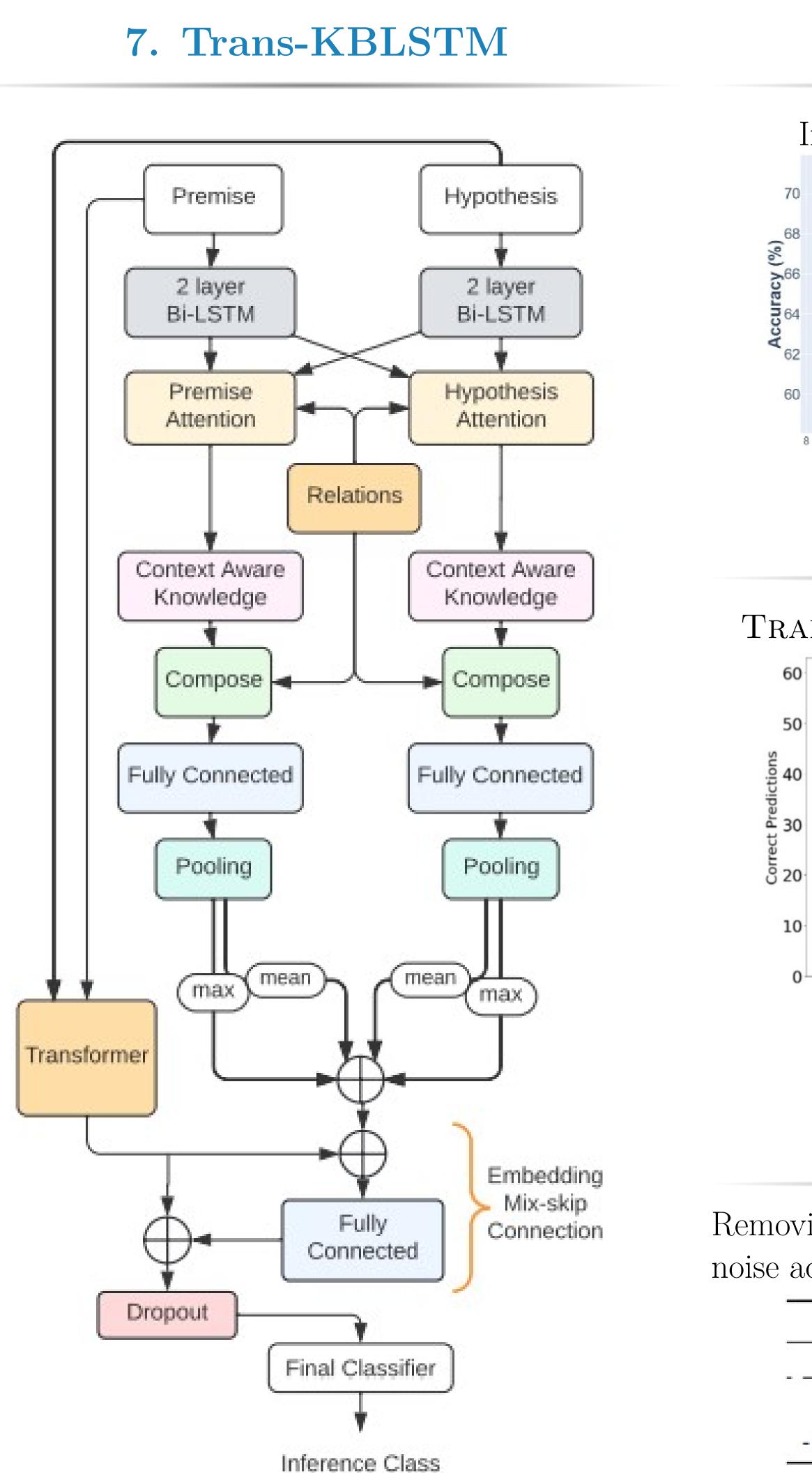


# 6. Knowledge Integration

• Word Pair External Knowledge Relations are not compatible with tokenized transformer inputs.



<sup>2</sup>BiLSTMs employ token level embeddings, thus complete word pair relations.



## 8. Full Supervision

TRANS-KBLSTM outperform other baselines.

Model	Dev	$\alpha 1$	$\alpha 2$	$\alpha 3$
w/o Knowledge	77.30	76.44	70.49	69.05
Tok-KTrans	78.17	76.19	70.75	69.77
KG Explicit	78.97	77.84	71.13	69.58
Trans-KBLSTM	79.92	79.62	72.10	70.21

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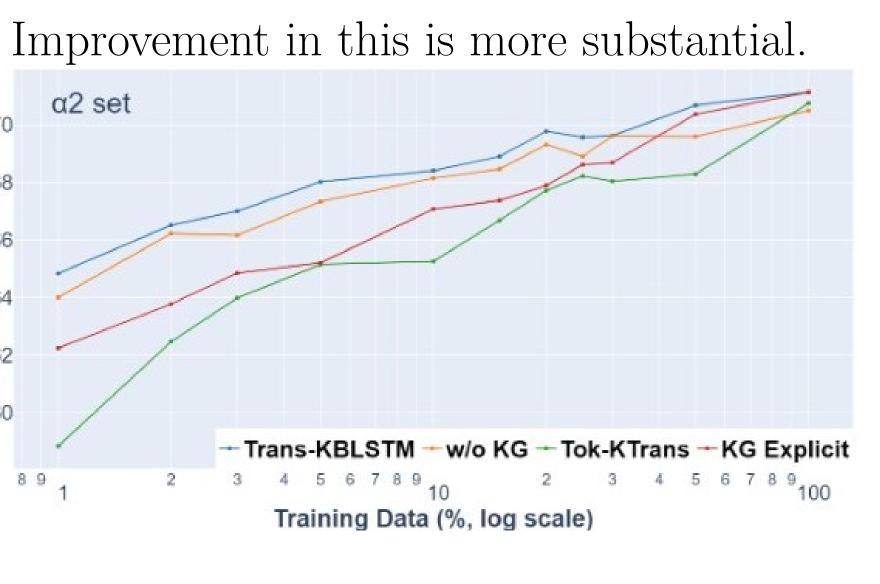
Joint training is better than Independent training

Ablations	Dev	$\alpha_1$	$\alpha_2$	$\alpha_3$
RoBERTaLARGE	77.30	76.44	70.49	69.05
+ KBLSTM (Independent)	79.22	78.38	71.00	69.22
+ KBLSTM (Joint Train)	79.92	79.62	72.10	70.21

Extensive analysis on InfoTabS dataset show TRANS-KBLSTM enhance performance.

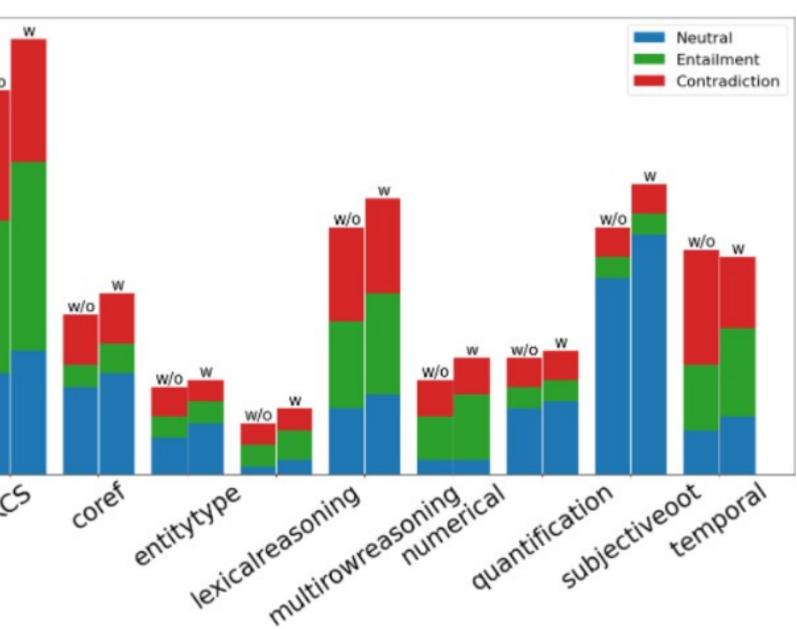


### 9. Limited Supervision



### **10. Reasoning Types**

### TRANS-KBLSTM improve all reasoning types.



# **11. Ablation Analysis**

Removing Skip connection and addition of random noise adversely affects model performance.

Ablations	Dev	$\alpha_1$	$\alpha_2$	$\alpha_3$
rans-KBLSTM	67.55	65.16	64.00	63.38
Skip Connect	65.72	62.83	60.00	61.55
- KB	60.44	61.88	56.94	55.55
B + Skip Connect)	60.11	61.50	55.94	57.38

Code: https://trans-kblstm.github.io