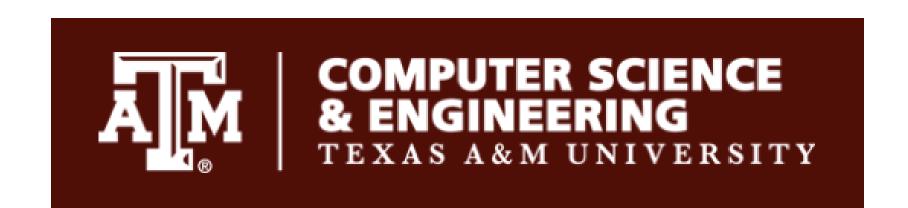
# **Explicitly Capturing Relations between Entity Mentions via Graph Neural Networks for Domain-specific Named Entity Recognition**



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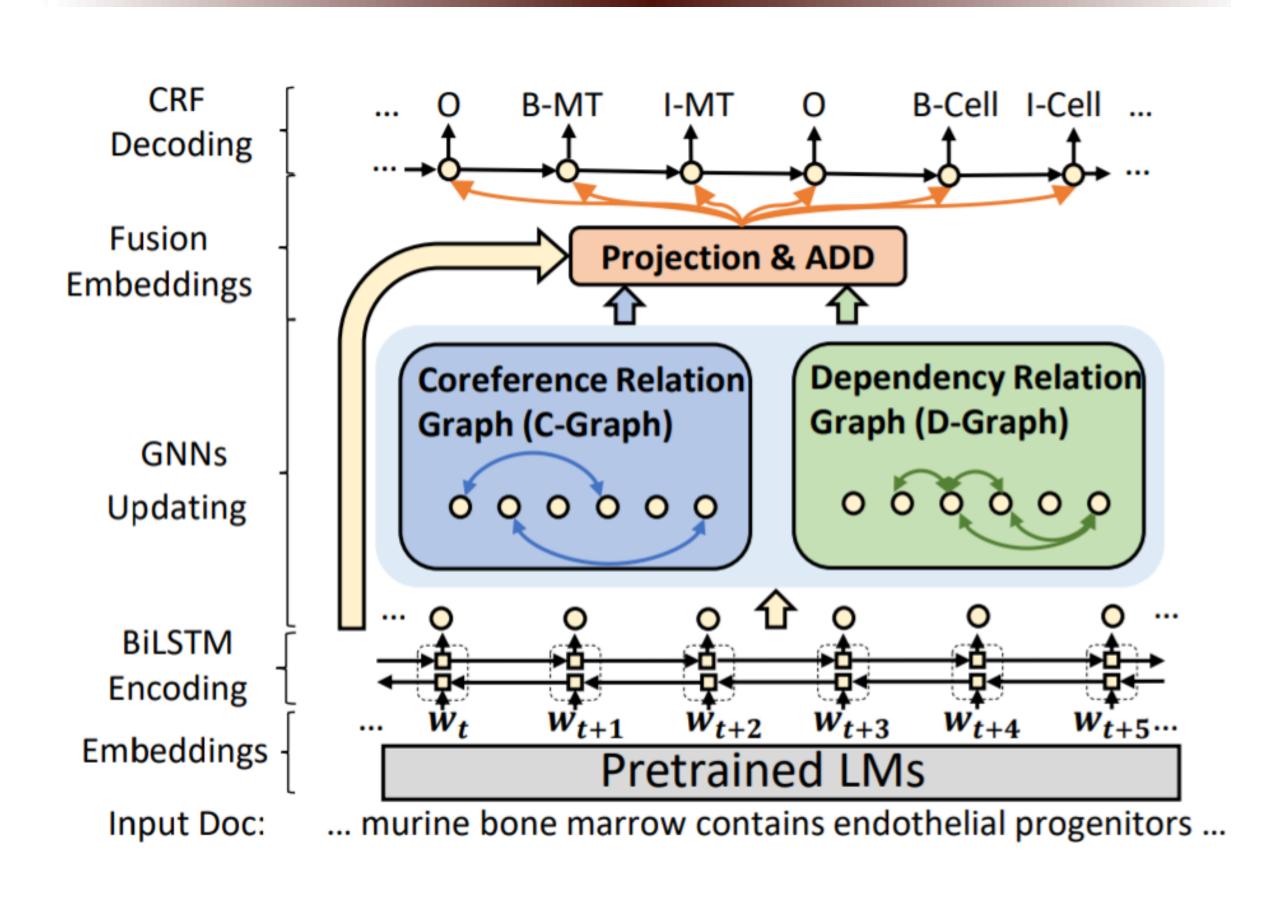
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#### Introduction

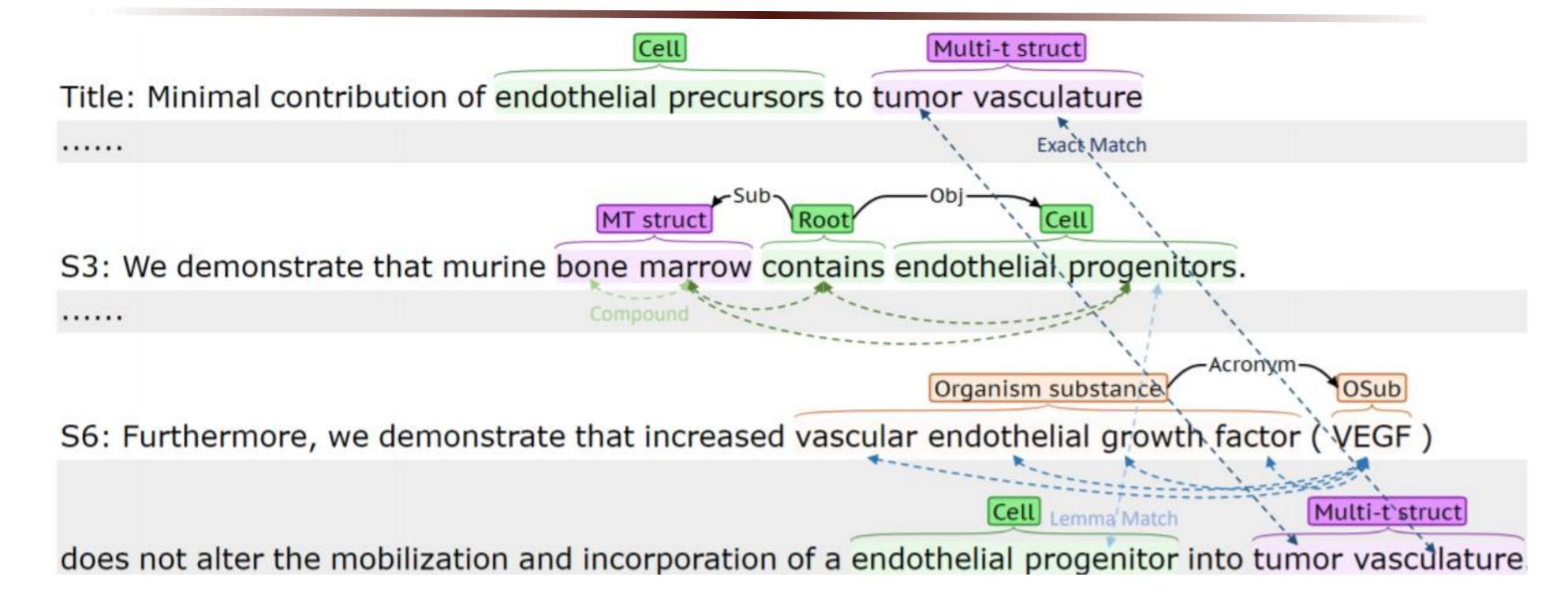
- The Named entity recognition (NER) performance is still moderate for **specialized domains** that have complicated contexts and jargonistic entity types.
- We hypothesize that the interactions of the related entity mentions (both document-level coreference and sentence-level dependency) will lead to better performance.
- Code for the system: https://github.com/brickee/EnRel-G

#### **Overall Architecture**



Use Graph Neural Networks to incorporate the relations between entity mentions.

## Relations between Entity Mentions



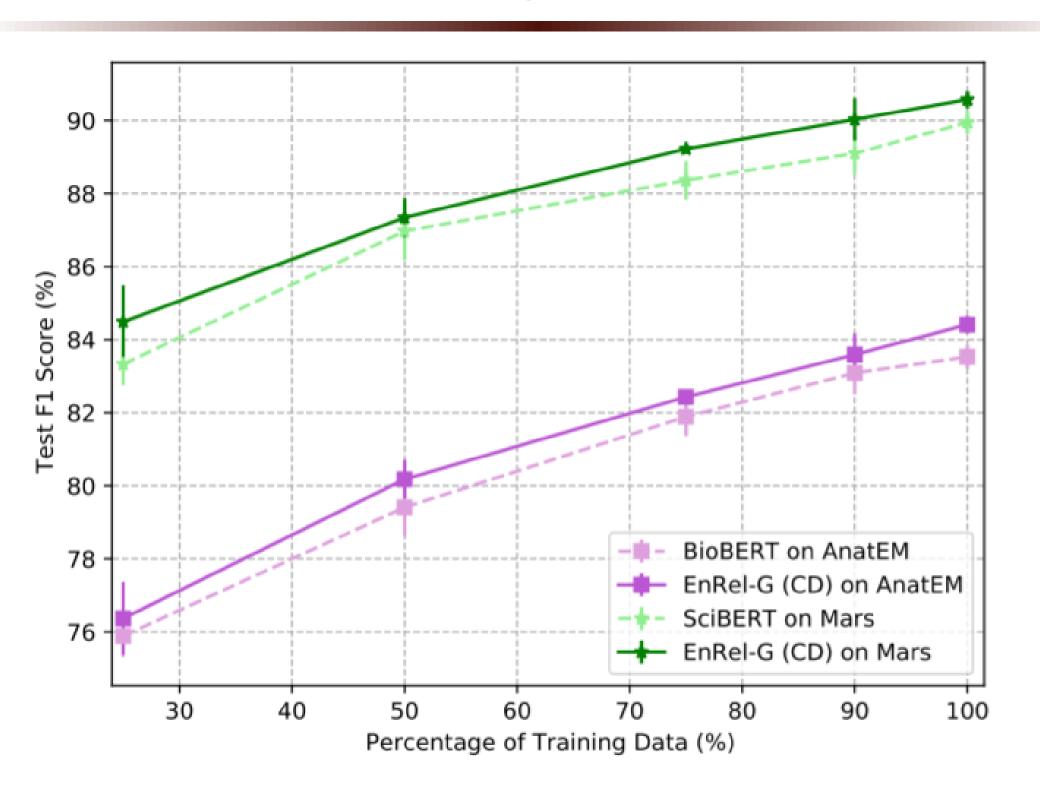
- The Coreference Relation Graph at document-level
- The Dependency Relation Graph at sentence-level

#### Results

Methods	Datasets	
	AnatEM	Mars
Wagstaff et al. (2018)	_	94.5 / 77.7 / 85.3
NCRF++	$83.40\pm0.34$ / $76.96\pm0.46$ / $80.05\pm0.12$	$91.28\pm1.08 / 80.57\pm0.55 / 85.59\pm0.23$
FLAIR	$81.07\pm0.29$ / $75.28\pm0.57$ / $78.06\pm0.39$	$90.67\pm1.02 / 81.45\pm1.41 / 85.81\pm0.62$
Pooled FLAIR	$82.11\pm0.50$ / $77.55\pm0.40$ / $79.76\pm0.34$	$87.79\pm1.31 / 86.57\pm1.10 / 87.17\pm0.17$
Tuning Bio/SciBERT	$83.94\pm0.40$ / $83.12\pm0.30$ / $83.53\pm0.32$	$90.93\pm0.66 / 88.99\pm1.61 / 89.95\pm0.64$
EnRel-G (C)	$84.65\pm0.67 / 83.69\pm0.31 / 84.17\pm0.41$	91.21±1.05 / <b>89.35</b> ±1.76 / 90.27±0.45
EnRel-G (D)	84.98±0.83 / 83.50±0.45 / 84.23±0.54	$92.66\pm1.16/88.03\pm1.46/90.29\pm0.53$
EnRel-G (CD)	$84.86\pm0.50$ / $83.96\pm0.32$ / $84.41\pm0.24$	$92.57\pm1.00 / 88.65\pm1.50 / 90.57\pm0.47$

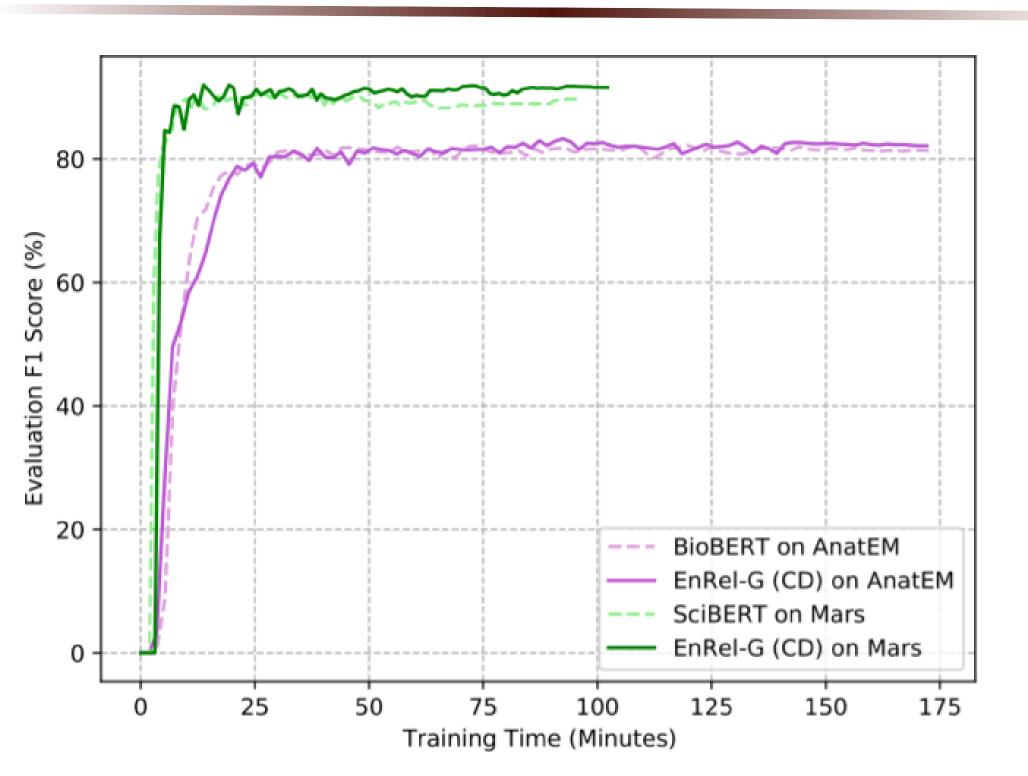
- Our system with both the global entity coreference and local dependency relations performs the best among all the systems.
- Either the coreference or dependency relations can help to improve the NER performance.

## Learning Curves



System is effective when only a tiny amount of labeled data is available

# **Computation Cost**



The lightweight approach does not increase the time cost compared to BERT models