

Zero-Shot Stance Detection: A Dataset and Model using Generalized Topic Representations

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Zero-Shot Stance Detection

Motivation

What opinions are implicitly conveyed in new articles?

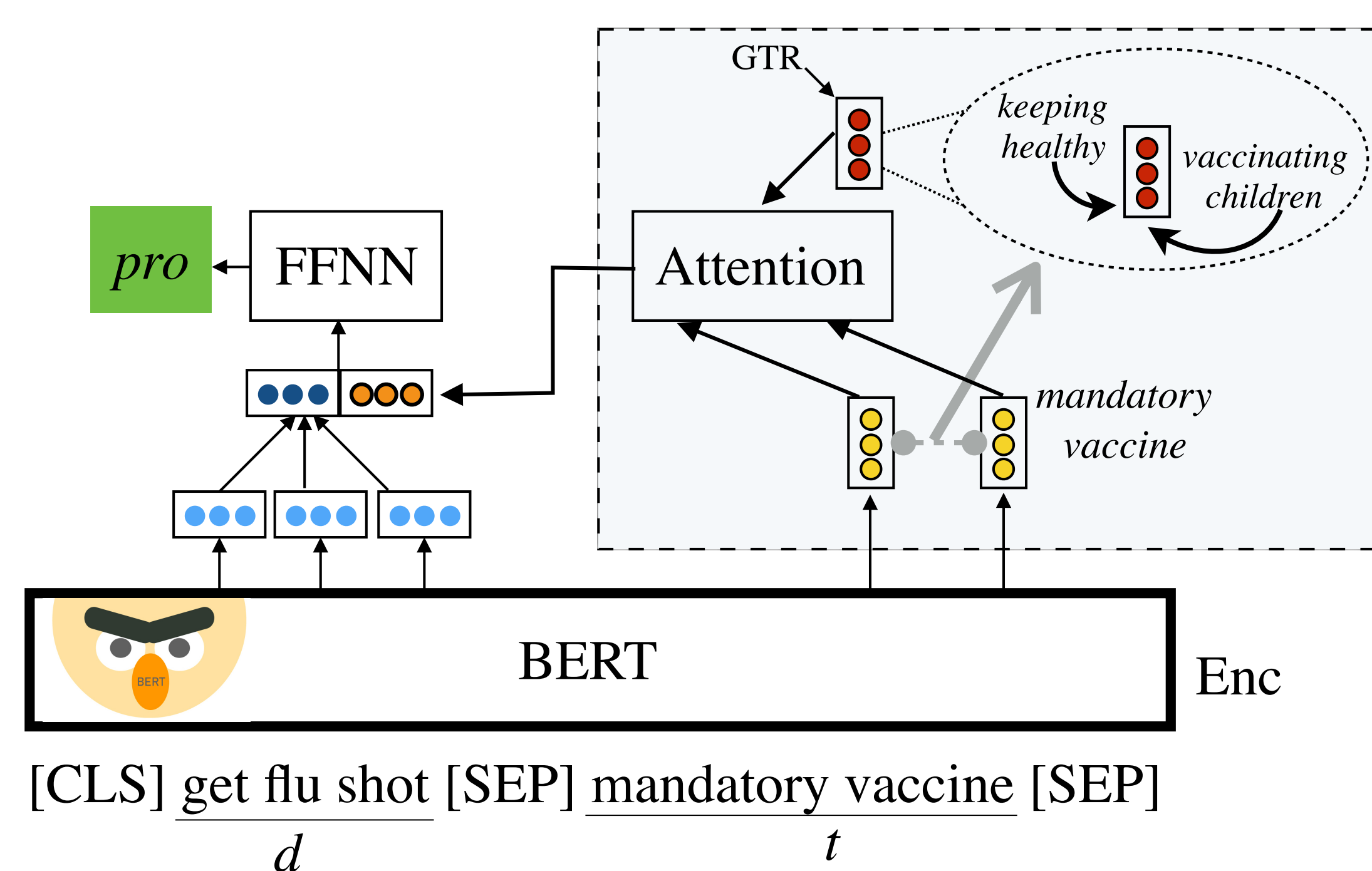
Topic: immigration **Stance:** against

Text: The jury's verdict will ensure that another **violent criminal alien** will be removed from **our community** for a very long period ...

Contributions:

- VAST: a *new dataset* for zero-shot and few-shot stance detection with many topics
- TGA Net: a *new model* using generalized topic representations
 - Improves on challenging linguistic phenomena
 - Relies less on sentiment cues

TGA Net



Goal: construct & use *implicit* relationships between seen and unseen topics

- unsupervised, no human knowledge!

Topic-Grouped Attention (TGA):

1. Cluster semantically similar topics
2. Assign *unseen* topic to closest cluster centroid (GTR)
3. Compute similarity between unseen topic and GTR with *attention*

VAST Data Collection

VARied Stance Topics (VAST)

Source: Comments from debate articles on *The New York Times*

	Comment	ARC Stance	Topic	ℓ
(1)	... Instead they have to work jobs (while their tax dollars are going to supporting illegal aliens) in order to put themselves through college [cont]	Immigration is really a problem	immigration a problem costs to american citizens	Con Con
(2)	Why should it be our job to help out the owners of the restaurants and bars? ... If they were paid a living wage ...[cont]	Not to tip	workplace living wage restaurant owners	Pro Con
(3)	...I like being able to access the internet about my health issues , and find I can talk with my doctors ... [cont]	Medical websites are healthful	medical websites home schoolers	Pro Neu

Statistics

<i>Ori Topic</i>	8010
<i>List Topic</i>	11531
<i>Neutral</i>	3984
Total Ex Topics	23525

Annotation

1. Heuristic-based topic extraction

- Existing Valence+topic label \rightarrow (1) and (3)
- categories from *NYT* \rightarrow (2)

2. AMT annotation

- Stance labeling $\rightarrow \ell$
- Topic correction \rightarrow (1)
- List other relevant topics \rightarrow (2)

Observations

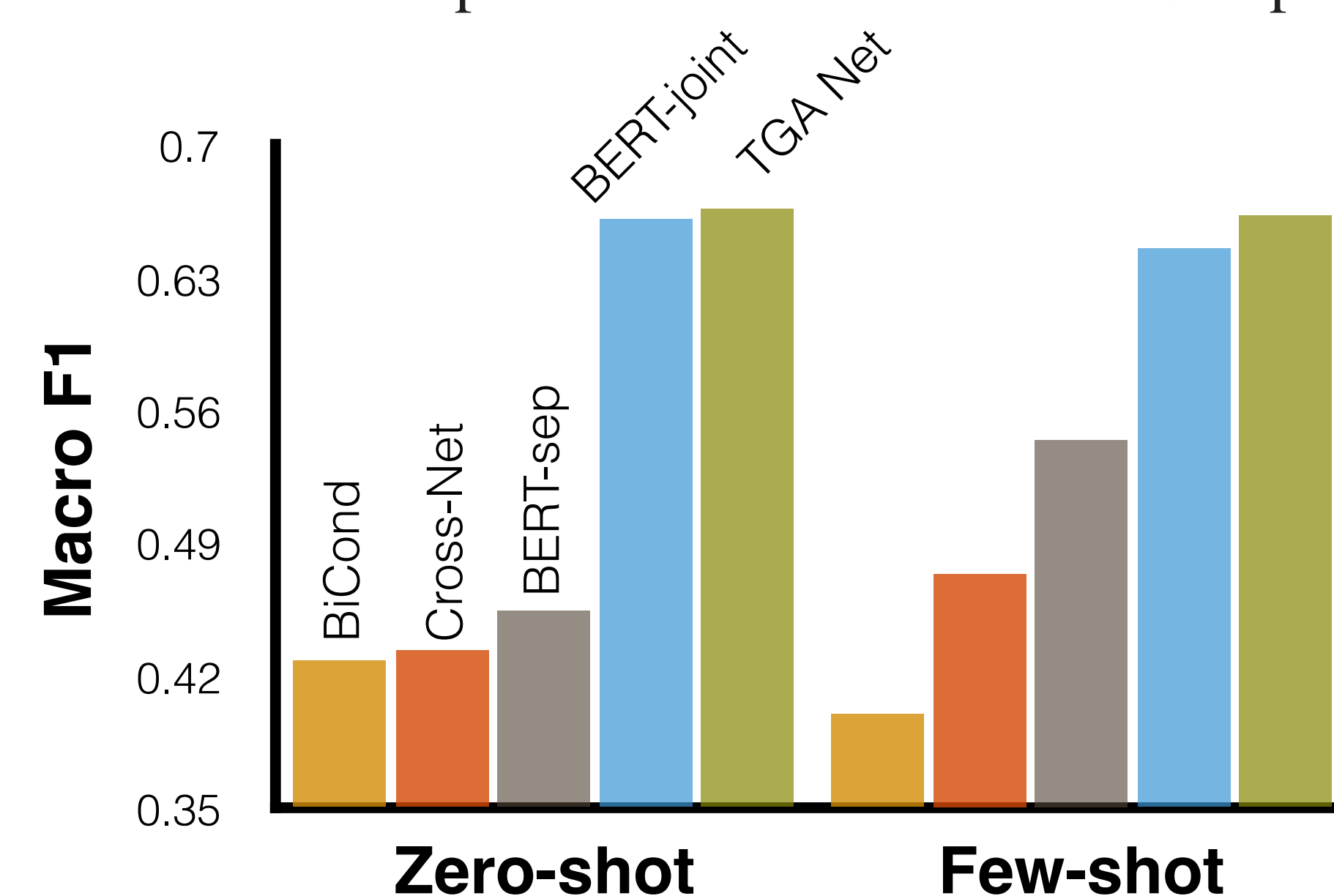
- Lexical variety in topics
- Multiple labels or topics per doc
- Topic corrections: $\sim 30\%$ of time

Evaluation and Analysis

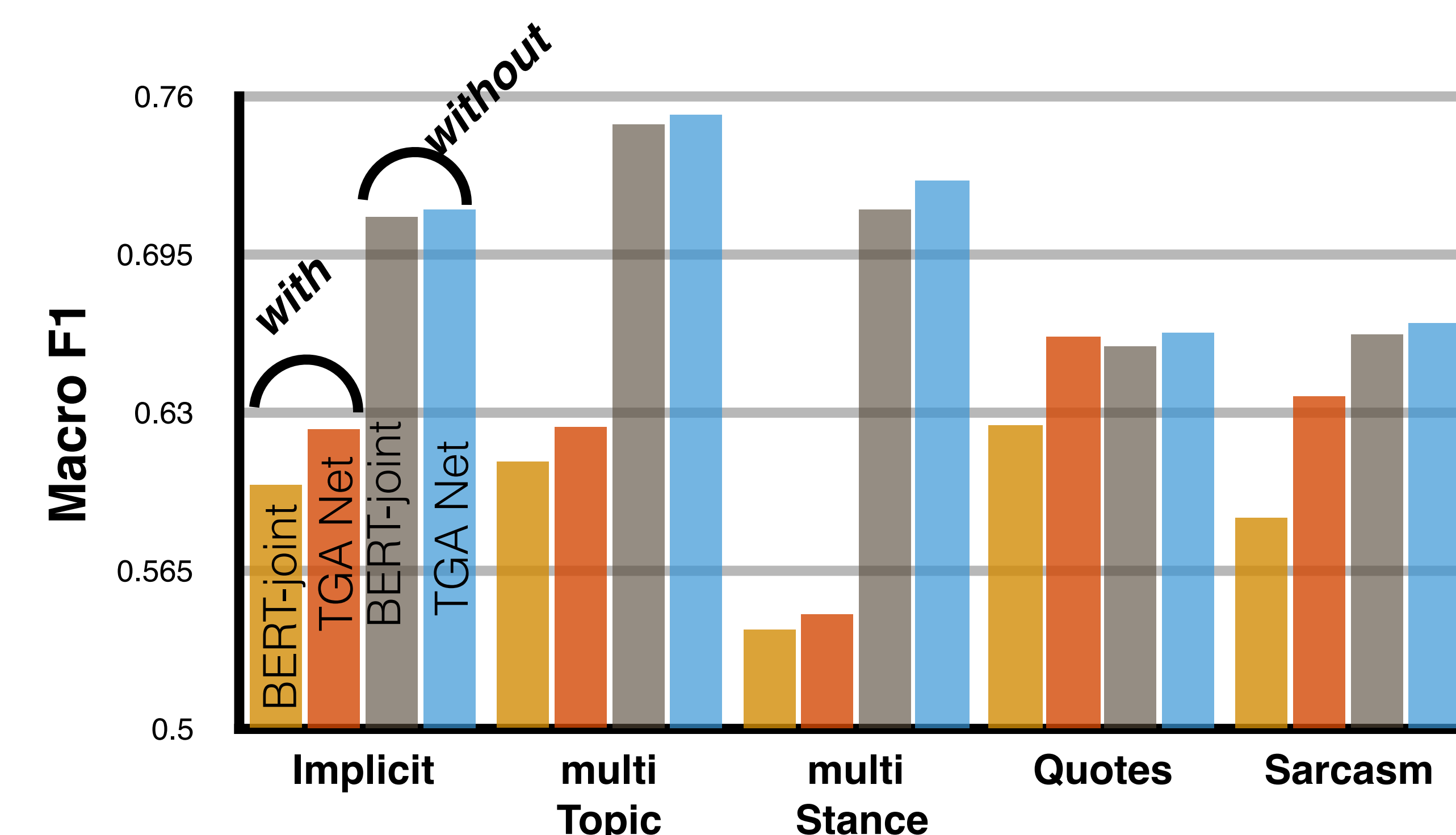
Stance Prediction

Zero-shot: 600 topics

Few-shot: 159 topics



Hard Phenomena



Dependence on Sentiment: swap sentiment words to confuse models

... debaters don't **strike(-)** \rightarrow **shine(+)** me as being anywhere near diverse in their perspectives on guns. Not one of the gun-gang cited any example of where a student with a gun saved someone from something **terrible(-)** \rightarrow **tremendous(+)** on their campuses. At **least(-)** the professor speaks up for **rationality(+)**.

guns

Sentiment words are **bold italicized**, removed words (~~struck out~~), and positive (green (+)) and negative (red (-)) sentiment words.