

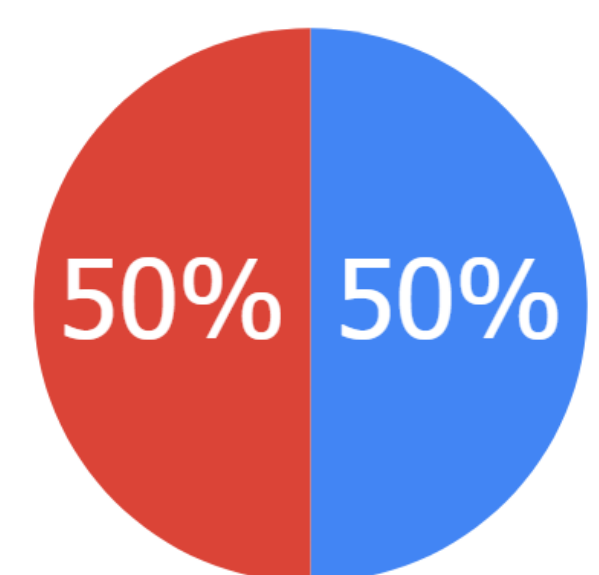
Motivation & Data

- **Hyperpartisan news are inherently propagandistic:**
 - Promote a particular non-objective point of view
 - Primarily used to influence public opinion or promote an agenda
- **Thus, we explore features known to work well for propaganda from A. Barron-Cedeno et al. (2019)**
- **Previous work:**
 - Determine factuality and bias of news sources rather than articles
 - Claim verification and stance detection

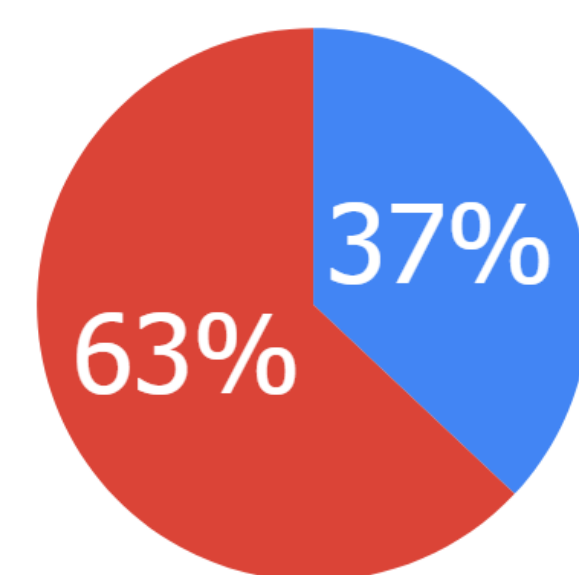


Dataset

Publisher Labels



Article Labels

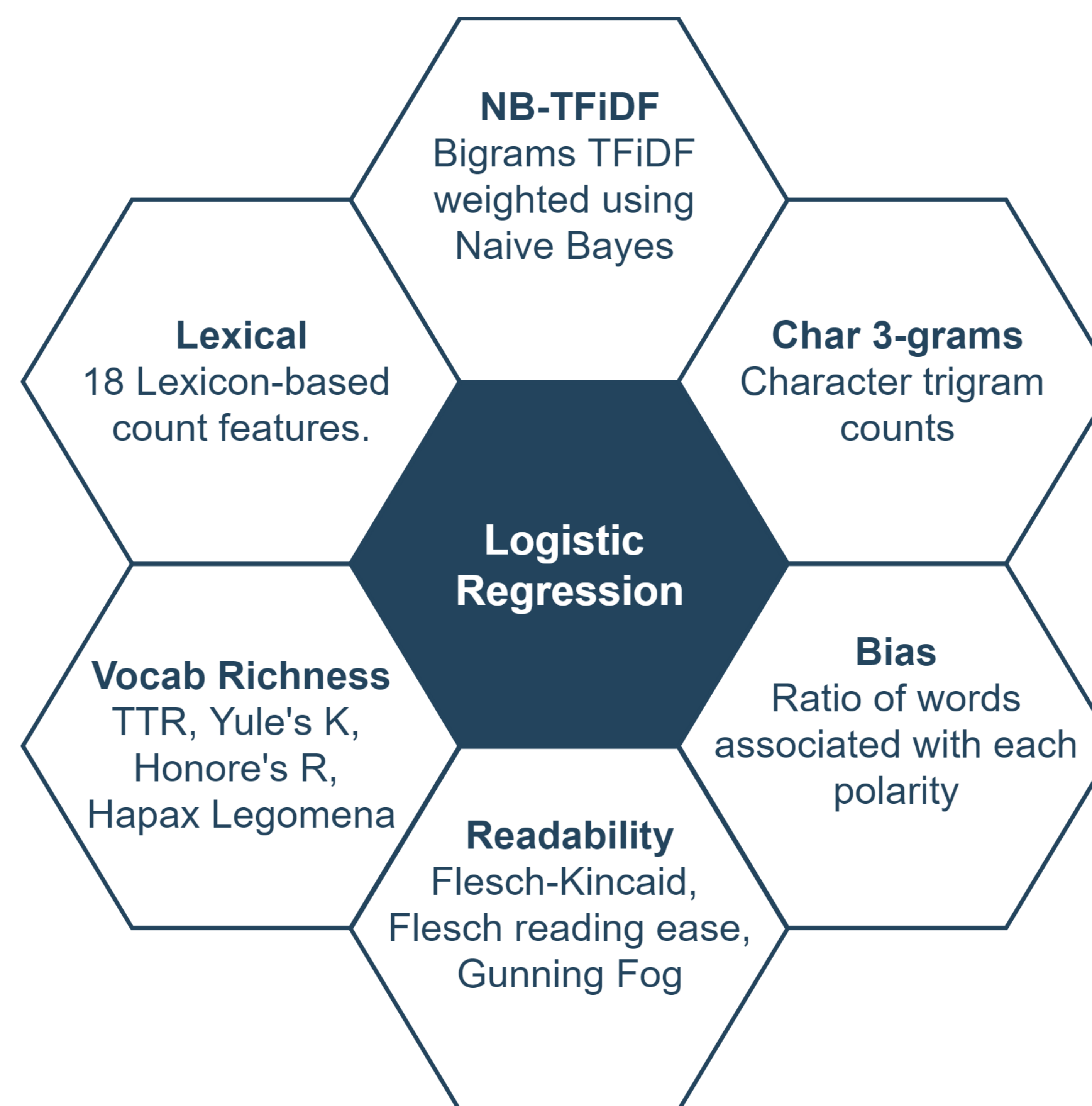


• partisan
• non-partisan

- **Publisher labels:**
 - 750k articles labeled using distant supervision.
 - Labels inferred from the publisher's bias
- **Article labels:**
 - 645 articles labeled through crowdsourcing
 - More relevant for this task, more fine-grained labels

System Description

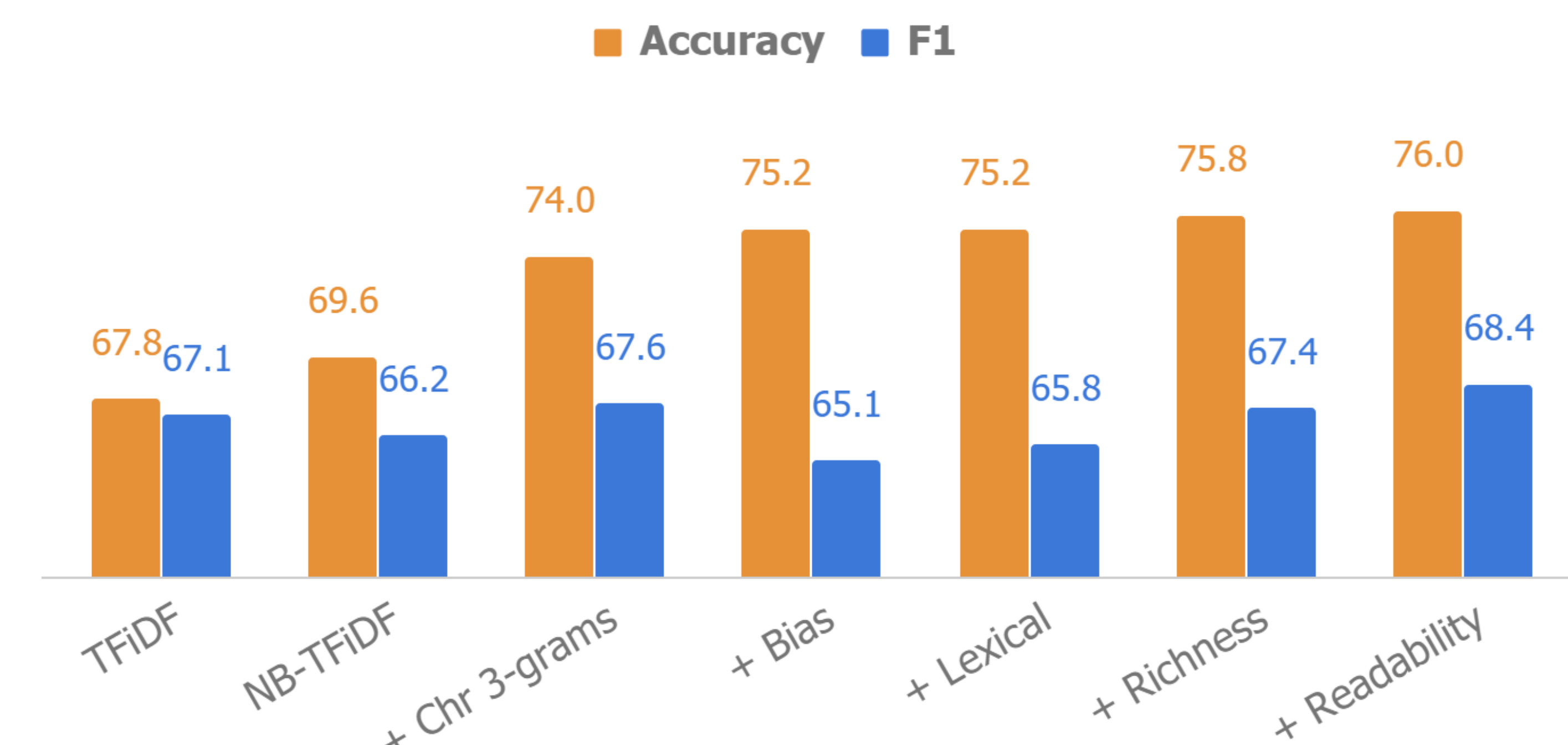
Features Used



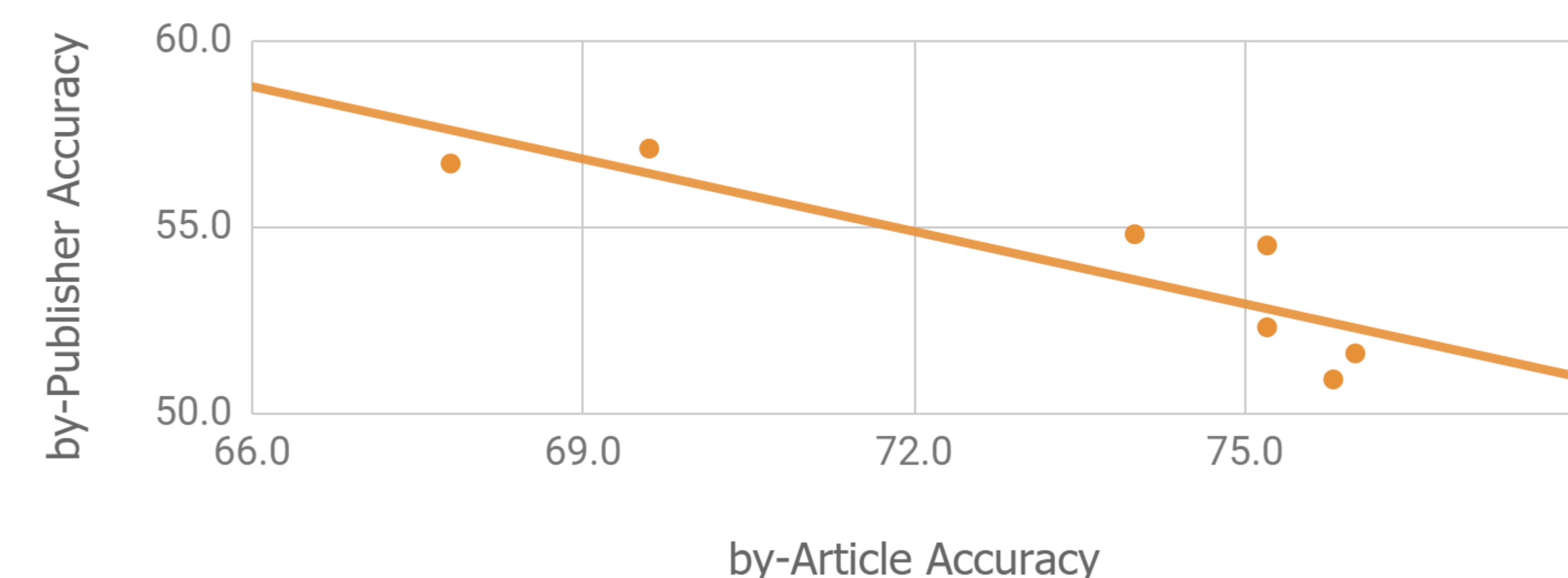
- **Training details:**
 - Train on the by-publisher data
 - Use the by-article data for validation
 - *Thus, avoid overfitting on the by-article data*
- **Model:**
 - logistic regression instead of more complex models
 - *Helps generalization since the training data is quite different from the test data*

Results

Labeled by-Article



Publisher vs Article Labels



- **Propaganda features classify hyperpartisan news:**
 - Significant improvements in accuracy for the by-article-labeled data
- **Modest improvements in F1 score:**
 - Adding propaganda features improves precision, but it also hurts recall.
 - However, recall is more important here.
- **Char trigrams are important:**
 - Largest improvement for a single added feature.
- **Publisher labels not ideal:**
 - By-Article accuracy negatively correlates with by-Publisher accuracy ($r = -0.87$)
 - Difficult to generalize from by-publisher data