Abstract

Rhetorical questions are frequently used in social media, aimed as capturing an audience’s attention or achieving a pragmatic goal, such as argumentation or persuasion. Although it has been observed that many of the uses of rhetorical questions are sarcastic, there is no large-scale computational work dedicated to distinguishing sarcastic from argumentative rhetorical questions, to our knowledge. To fill this gap, we expand on the only small existing dataset to include 2.5K rhetorical questions in debate forums. We use an LSTM model to distinguish between the classes, beating previous baselines with up to 0.75F for SARCASTIC and 0.76F for ARGUMENTATIVE, then analyze characteristic linguistic categories by class.

1 Introduction

Rhetorical questions (RQs) are prevalent in argumentative and persuasive discourse (Ilie, 1994). RQs are syntactically formulated as a question, but function as an indirect assertion (Frank, 1990), so previous work on RQs focuses on their formal semantic properties (Han, 1997).

Interestingly, while Gibbs (2000) and others mention that RQs are often used ironically (Schaffer, 2005; Ilie, 1994), there is no previous work that focuses on distinguishing sarcastic from argumentative (or other) uses of RQs at scale. Gibbs’ (2000) taxonomy defining a non-unitary basis for irony is based on five categories: sarcasm, hyperbole, understatement, jocularity, and finally rhetorical questions. Likewise, Ilie (1994) points out that RQs and their responses may also be used to project irony, humor, and sarcasm.

In previous work, we use Gibbs’ (2000) taxonomy of sarcasm to build a corpus of three different forms of sarcasm (general, hyperbole, and RQs) (Oraby et al., 2016). The corpus of RQs in debate uses a heuristic to identify RQs depending on their position within a post in a dialog. The final RQ dataset includes 1,702 posts balanced between the SARCASTIC and ARGUMENTATIVE classes (851 posts per class).

We choose to focus specifically on characterizing the use of RQs in online debate. Our contributions in this work are:

- Development of a larger dedicated corpus of 2,496 RQs in debate based on a heuristic we define in previous work (Oraby et al., 2016).
- Presenting classification results using an LSTM model to classify SARCASTIC vs ARGUMENTATIVE RQs, beating the baseline.
- Analysis of the linguistic categories of RQs in debate, highlighting stylistic differences.

2 Data Collection

In order to do an in-depth study of RQs in debate, we need a reasonably large set of RQs. We consider debate posts from the Internet Argument Corpus (IAC) 2.0 (Abbott et al., 2016), and the corpus of 1,702 RQs from our previous work (Oraby et al., 2016).¹

To gather the data, we define and apply a heuristic that gathers RQs based on the simplifying assumption that a question in a post is an RQ if the question occurs in the middle of the speaker’s turn, and if it is followed by a statement, indicated by a period or exclamation point (Oraby et al., 2016).

Thus, the assumption is that when speakers follow a question with a declarative statement, they

¹The only other related corpus that includes RQ annotations is the set of 583 RQs in the Switchboard-DAMSL dialog corpus (Rohde, 2006; Jurafsky et al., 1997).
Table 1: Supervised Learning Results for RQs in Debate Forums

<table>
<thead>
<tr>
<th>Features</th>
<th>SARC</th>
<th>ARGUMENTATIVE</th>
<th>Average</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>P</td>
<td>R</td>
<td>F</td>
</tr>
<tr>
<td>W2V (on RQ)</td>
<td>0.78</td>
<td>0.62</td>
<td>0.68</td>
</tr>
<tr>
<td>W2V+LIWC (on RQ)</td>
<td>0.76</td>
<td>0.68</td>
<td>0.72</td>
</tr>
<tr>
<td>W2V+LIWC (on Pre+RQ)</td>
<td>0.81</td>
<td>0.60</td>
<td>0.69</td>
</tr>
<tr>
<td>W2V+LIWC (on RQ+Post)</td>
<td>0.74</td>
<td>0.76</td>
<td>0.75</td>
</tr>
<tr>
<td>W2V+LIWC (on Full Text)</td>
<td>0.76</td>
<td>0.67</td>
<td>0.71</td>
</tr>
</tbody>
</table>

3 Results and Analysis

We conduct supervised learning experiments using an LSTM model from Keras (Chollet, 2015) with TensorFlow backend (Abadi et al., 2016), dividing our data randomly into 80% training and 20% test and doing a grid-search on our training data for parameter tuning. For features, we use Google News-trained Word2Vec (Lowe et al., 2015). We experiment with adding in Linguistic Inquiry and Word Count (LIWC) scores (Pennebaker et al., 2001), and training with different levels of post-level context: pre (the RQ and what precedes it), RQ (just the RQ), and post (the RQ and what follows it).²

Table 1 shows our experimental results for each class, as well as the averages. Results increase over our simple W2V baseline for both SARC and ARGUE when using LIWC categories, with best results coming from training using the post context for SARC (p < 0.05 compared to the baseline), and on the pre for ARGUE. It is frequent that in the SARC posts, the speaker ends with a sharp remark or interjection like “gasp!” or an emoticon like wink :) or rolleys 8-). The ARGUE posts are often nestled within sequences of questions or other RQ pairs, seeking to inform. Our top results of 0.75F SARC and 0.76F ARGUE improve on those from our previous work (0.70 for SARC and 0.71 for NOT-SARC) with an SVM classifier and n-gram features using 10-fold cross validation on our original smaller dataset (Oraby et al., 2016).

We investigate the data to better understand the role context plays in the classification. An example RQ in context is shown in Table 2, divided into the pre, RQ and post. We observe that although the RQ itself may not appear sarcastic on its own, the post context makes the sarcasm much more pronounced. A qualitative analysis of the SARC vs. ARGUMENTATIVE data shows that sarcastic RQs are often followed by short statements that serve to mock, whereas the argumentative pairs are used to structure an argument or draw attention to a point.

Table 3 shows examples of the top LIWC features that surface for both classes in our experiments.³ In Row 1, we observe that 2nd person mentions are frequent in the sarcastic debate, which aligns with previous work citing that having an “identifiable victim of irony, sarcasm, or both” is heuristic for ironic interpretation (Kreuz and Glucksberg, 1989; Katz, 1996; Gibbs, 2000). Informal words and more “verbal speech style” non-fluencies, including exclamations and social media slang (Rows 2-4), are also characteristic features.

²The length of the pre and post context vary based on the length of the post, which range between 10-150 words.

³We find the top categories empirically by using learned feature weights (FW) on a subset of the data.
<table>
<thead>
<tr>
<th>#</th>
<th>FW</th>
<th>Feature</th>
<th>Example</th>
</tr>
</thead>
</table>
| 1  | 13.5 | 2
Person | Do you ever read headers? You got a mouth on you as big as grand canyon. |
| 2  | 10.9 | Informal | The hate you’re spewing is palpable, yet you can’t even see that can you? Hypocrites, ya gotta luv em. |
| 3  | 10.4 | Exclamation | Force the children to learn science? How obscene!! |
| 4  | 5.56 | Netspeak | To make fun of my title? lol, how that stings... |

### Argumentative

<table>
<thead>
<tr>
<th>#</th>
<th>FW</th>
<th>Feature</th>
<th>Example</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>7.29</td>
<td>Interrog.</td>
<td>How do you know it’s the truth? If it were definitive, I’d be more inclined [...]</td>
</tr>
</tbody>
</table>
| 6  | 6.30 | 3
Person Plural | what’s the difference? both are imposing their ideologies |
| 7  | 3.84 | Differ. | if you can’t, why do you keep disputing me? like i said [...] |
| 8  | 3.35 | Health | When will the people press congress to take up abortion? It’s the job of congress [...] |

Table 3: LIWC Categories in Debate

For argumentative debate, we observe that interrogatives as the most distinguishing features, where questions are phrases with who, what, where, when, why, how. “Differentiation” is a strong feature of argumentative forums (Row 7), relating to comparative words such as “against” and “alternatively”, as well as technical jargon in Row 8 (genre-specific topics, such as abortion).

### 4 Conclusion

To our knowledge, this is the first study dedicated specifically to sarcasm in RQs. Our results show that sarcastic and argumentative uses of RQs are highly distinguishable using linguistic features and context. We show that SARCAS-TIC RQs are often informal and include 2
Person references, while ARGUMENTATIVE RQs contain technical jargon and interrogatives. Our future work includes exploring RQs in different domains, using more robust models, and developing more context features.

### References


Hannah Rohde. 2006. Rhetorical questions as redundant interrogatives. Department of Linguistics, UCSD.
