Identifying and Characterizing Logical Fallacies

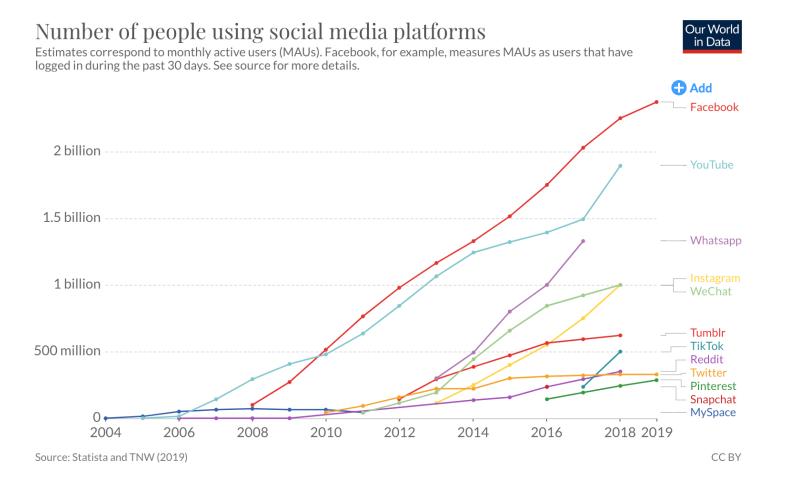
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Exponential Rise in Social Media Usage





Standard Format of Arguments

```
Since (premise),
which is a conclusion supported by (subpremise),
and (premise),
which is a conclusion supported by (subpremise),
and (premise),
[and (implicit premise)]
and (rebuttal premise),
Therefore, (conclusion).
```



A well-formed structure



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- Premises that are relevant to the truth of the conclusion



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- Premises that are acceptable to a reasonable person



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But, human reasoning is often affected by cognitive biases.



Violate one of the criteria of a good argument

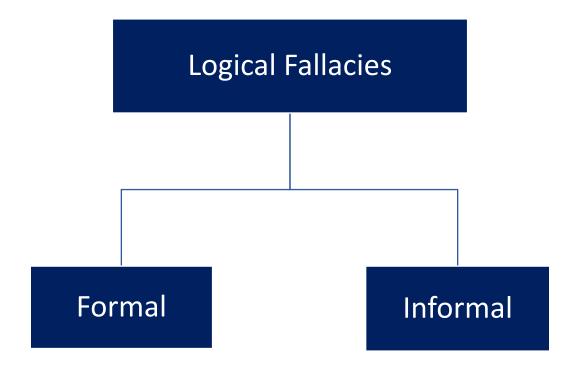


- Violate one of the criteria of a good argument
- Seem convincing, but their premises doesn't prove or disprove the conclusion.



- Violate one of the criteria of a good argument
- Seem convincing, but their premises doesn't prove or disprove the conclusion.
- Lead to disagreements, conflicts, endless debates, and a lack of consensus.







Formal Logical Fallacies

• Flaw in logical structure



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- May have true premises, but false conclusion



Formal Logical Fallacies

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• Example:

- Denying the antecedent
 - If *A* is true, then *B* is true.
 - A is false.
 - Therefore, *B* is false.

- If I am Indian, then I am Asian.
- I am not Indian.
- Therefore, I am not Asian.



Informal Logical Fallacies

• Incorrect argument in Natural Language



Informal Logical Fallacies

- Incorrect argument in Natural Language
- Fallacious not just because of the logical form, but also due to content and context.



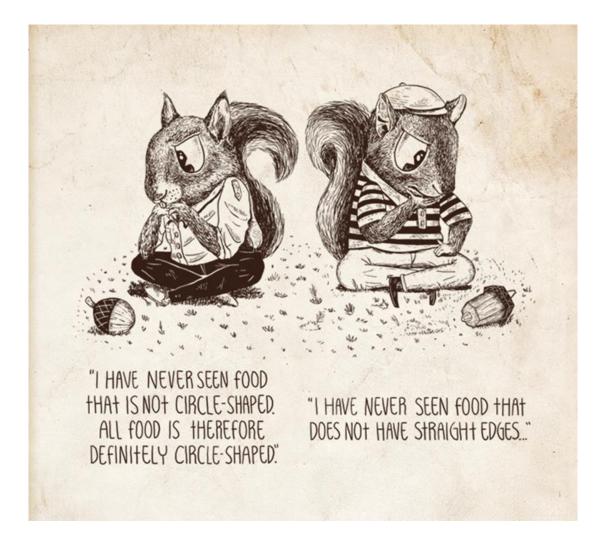
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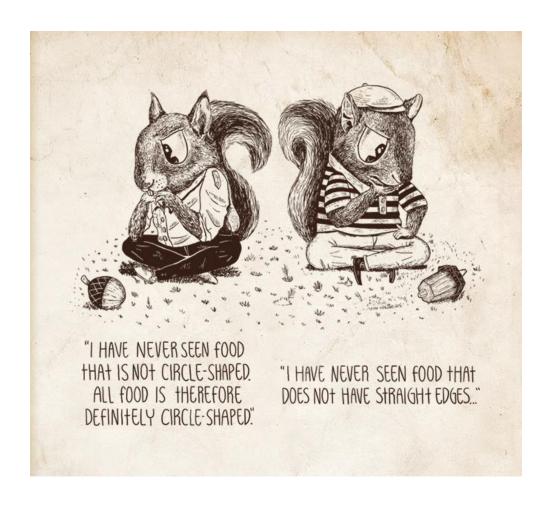
• Example:

- False Dilemma
 - A claim presenting only two options or sides when there are many options or sides.
 - You're either for the war or against the troops.



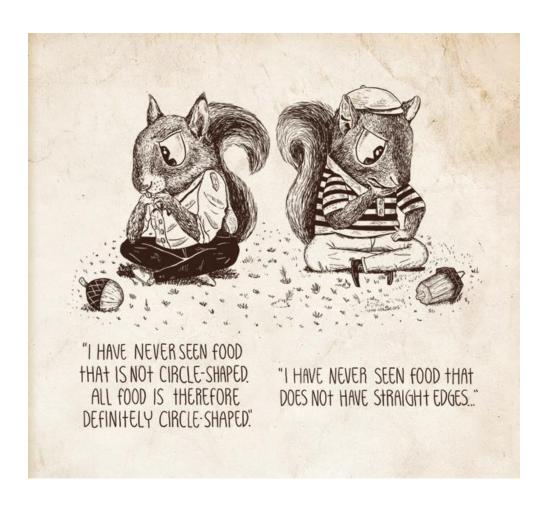






Faulty Generalization

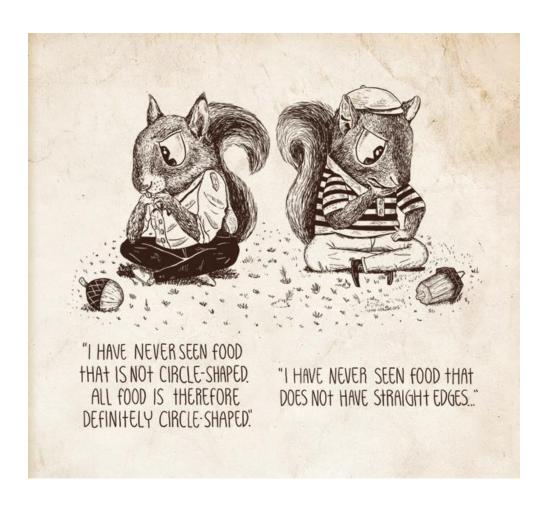




Faulty Generalization

Definition: an argument applies a belief to a large population without having a large enough sample to do so.



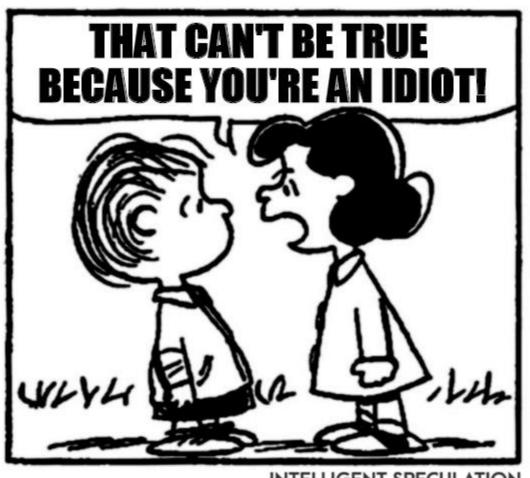


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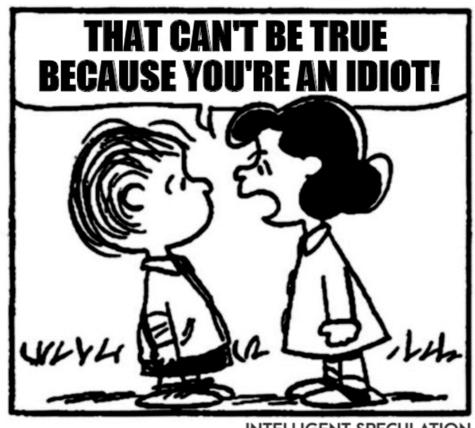
Example: I ordered chilly-mushroom for lunch. It tasted so bad, I will never order it again in my life.







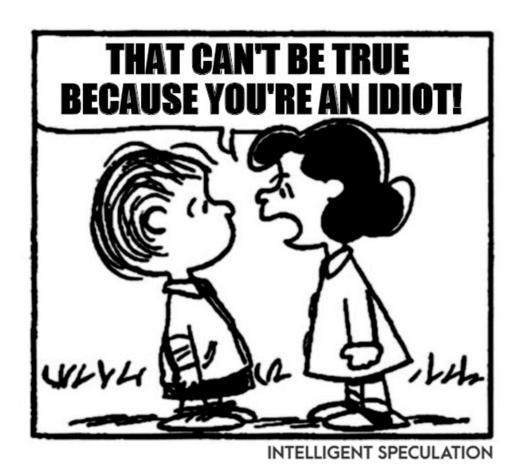




Ad Hominem



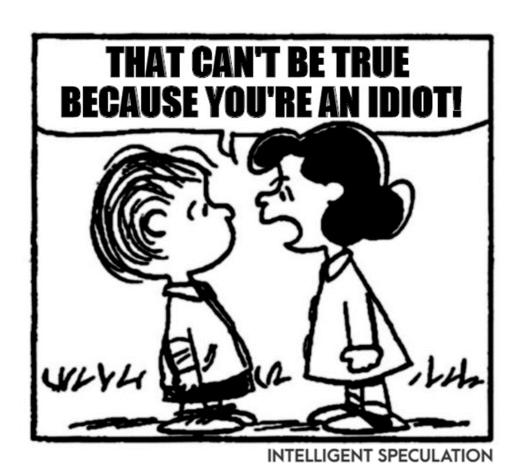




Ad Hominem

Definition: Speaker trying to argue the opposing view on a topic makes claims against the other speaker instead of the position they are maintaining.





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Definition: Speaker trying to argue the opposing view on a topic makes claims against the other speaker instead of the position they are maintaining.

Example: He supports Democrats. Hence, don't believe what he says.



Experiments and Results

Experiment 1
Logical Fallacy Detection in CreateDebate Forum



Social networking website hosting debates since February 2008.



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- Divided into 14 topical forums Politics, Religion, Science, etc.



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- Weak moderation policy
 - Only the debate creator can moderate



- Social networking website hosting debates since February 2008.
- Divided into 14 topical forums Politics, Religion, Science, etc.
- Weak moderation policy
 - Only the debate creator can moderate
- Perfect testbed for analyzing usage of informal logical fallacies



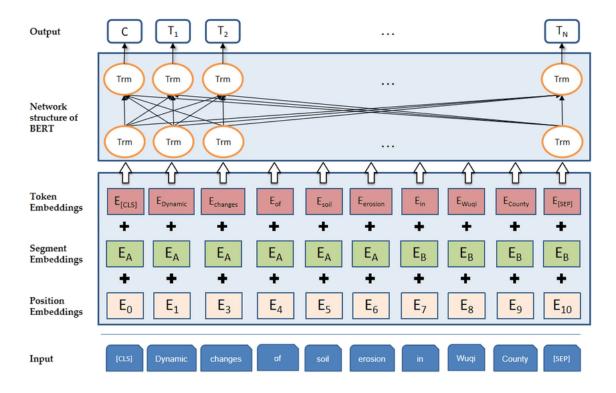
Topic	# Posts	# Comments	# Users
Politics	10,434	119,850	7,686
Religion	2,841	77,418	4,563
World News	2,008	27,418	3,622
Science	1,276	20,691	2,837
Law	759	11,016	1,436
Technology	909	8,421	2,674
Total	18,227	264,814	14,961

Table 3.1: Basic statistics of our collected CreateDebate dataset. The first post in our dataset was posted on February 20, 2008 and the last post was updated on November 24, 2021 (Patel et al., 2023).



Fine-tuning BERT

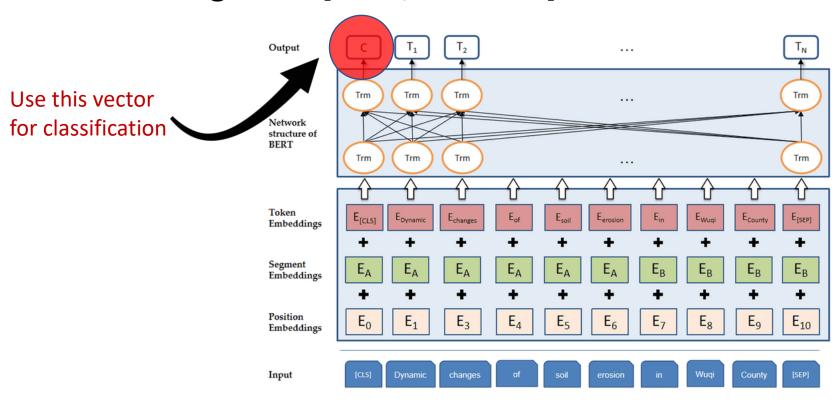
• We are using BERT [base, uncased] model





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Fine-tuning BERT

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- Fine-tune the model on LOGIC dataset



• Dataset published in (Jin et al., 2022).



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- Contains 2,449 examples of logical fallacies
 - Training partition: 1849
 - Validation partition: 300
 - Test partition: 300
- Examples categorized in 13 classes



Class Labels in LOGIC dataset

Class Label	Percentage (%)
Faulty Generalization	18.01
Ad Hominem	12.33
Ad Populum	9.47
False Causality	8.82
Circular Claim	6.98
Appeal to Emotion	6.82
Fallacy of Relevance	6.61
Deductive Fallacy	6.21
Intentional Fallacy	5.84
Fallacy of Extension	5.76
False Dilemma	5.76
Fallacy of Credibility	5.39
Equivocation	2.00

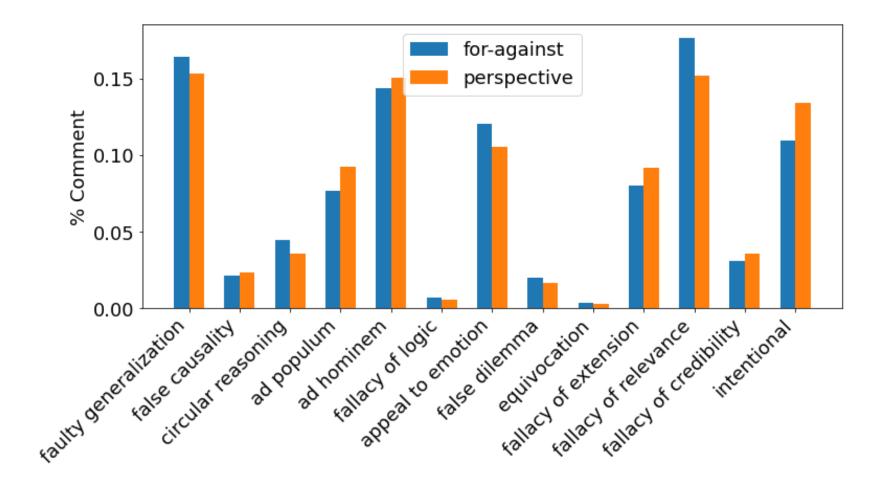


Fine-tuning BERT

- We are using BERT [base, uncased] model
- Fine-tune the model on LOGIC dataset
- Use the fine-tuned model to classify comment in Politics forum of CreateDebate



Distribution of Logical Fallacies in CreateDebate's Politics Forum





Experiments and Results

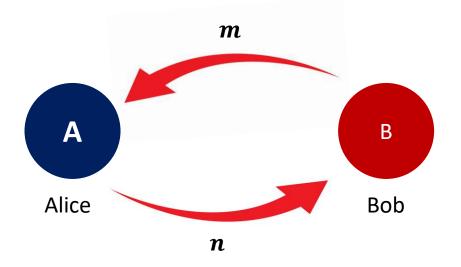
Experiment 2
Studies on CreateDebate's User Network



• User network can be built using two approaches:



- User network can be built using two approaches:
 - Nested structure of comments in a post



Alice directly replied to Bob n times Bob directly replied to Alice m times



- User network can be built using two approaches:
 - Profile pages of users







 CreateDebate hosts two types of debates: For-against and Perspective.



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For-against debates

• Users take a stand on a particular topic and argue in favor or against it



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For-against debates

Users take a stand on a particular topic and argue in favor or against it

Perspective debates

 Users share their viewpoints on a topic without necessarily taking a stance for or against it.



Topic	For-Against	Perspective	For-Against	Perspective
	$\mathbf{Posts}~\%$	$\mathbf{Posts}~\%$	${\bf Comments}~\%$	Comments %
Politics	48%	52%	68%	$\overline{32\%}$
Religion	60%	40%	76%	24%
World News	58%	42%	73%	27%
Science	55%	45%	74%	26%
Law	66%	34%	82%	18%
Technology	54%	46%	79%	21%

Table 3.2: Distribution of For-Against debates and Perspective debates across different forums.



We use both types of networks for our study

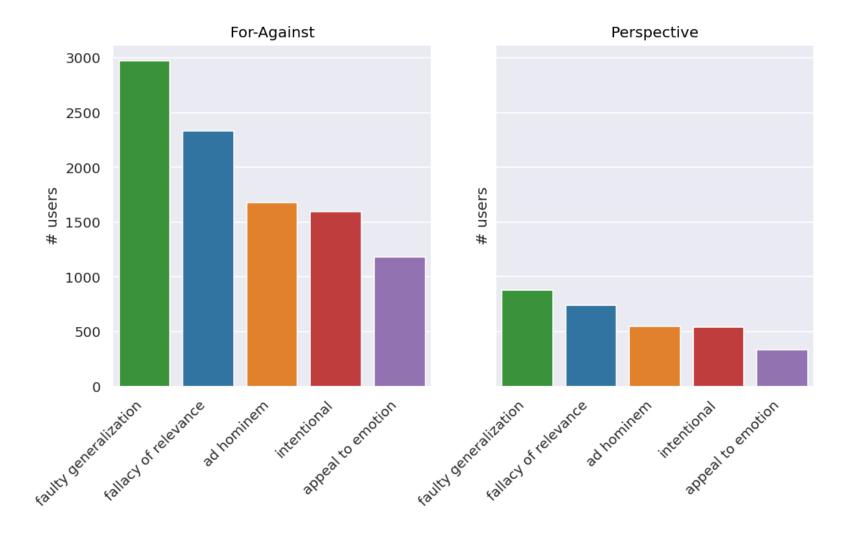


- We use both types of networks for our study
- As CreateDebate hosts two types of debates
 - For-against user network
 - Perspective user network



Number of users posting fallacious comments

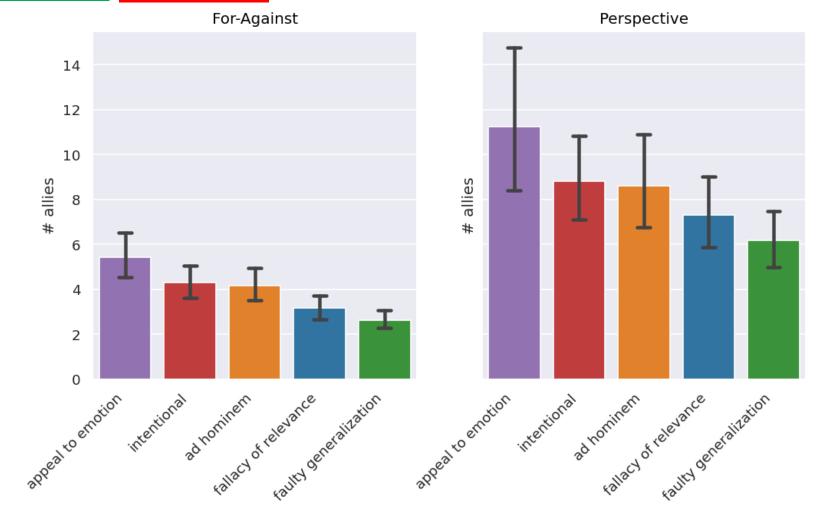
Order is preserved 3x decrease





Number of Allies

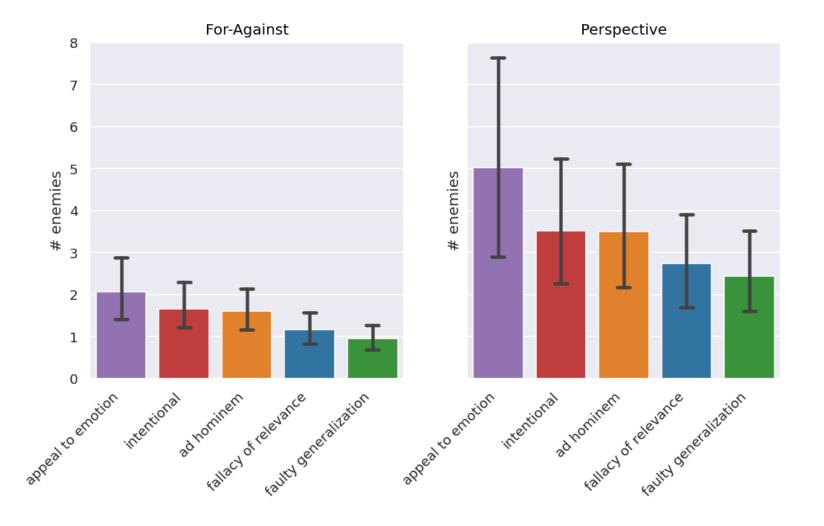
Order is preserved 2x increase





Number of Enemies

Order is preserved 2.5x increase

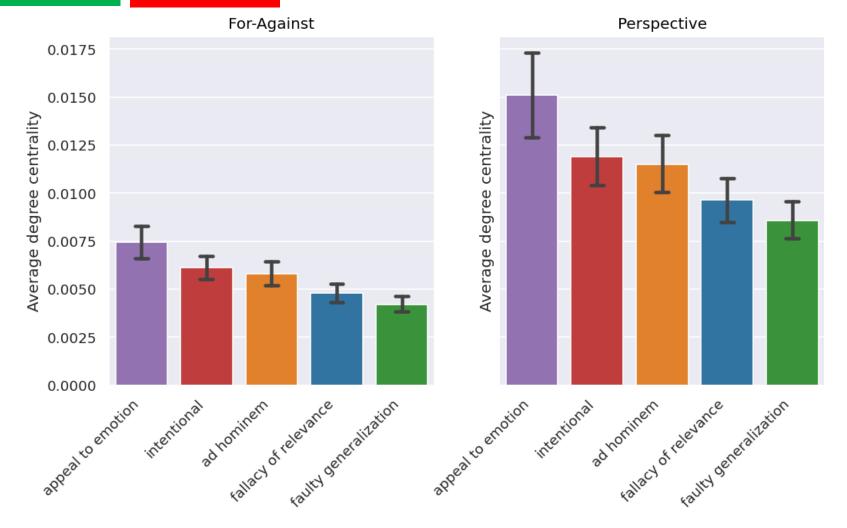




Degree centrality

Order is preserved 2x increase

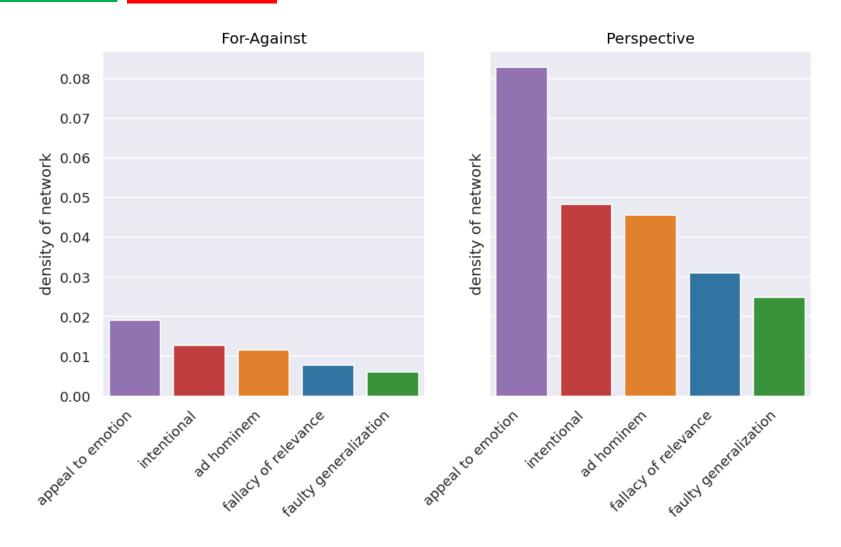
The degree centrality values are normalized by dividing by the maximum possible degree in a simple graph n-1 where n is the number of nodes in G.





Density of network $d = \frac{m}{n(n-1)}$

Order is preserved 4x increase





Experiments and Results

Experiment 3
Linguistic Study of Logical Fallacies in CreateDebate



Analyzes grammatical structure of a sentence

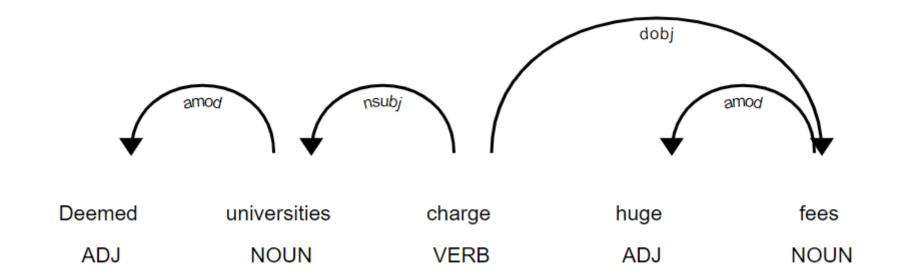


- Analyzes grammatical structure of a sentence
- Determines relationships between words



- Analyzes grammatical structure of a sentence
- Determines relationships between words
- Dependency tree is built for every comment in Politics forum







• Published in (Fast et al., 2016).



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- Tool for analyzing text across lexical categories



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- Has 194 built-in pre-validated topical and emotional categories



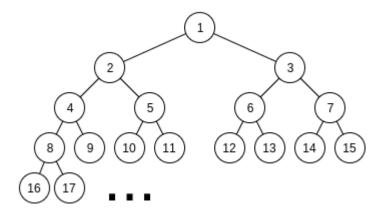
- Published in (Fast et al., 2016).
- Tool for analyzing text across lexical categories
- Has 194 built-in pre-validated topical and emotional categories
- Each category has on average 83 words.



social media	war	violence	technology
facebook	attack	hurt	ipad
instagram	battlefield	break	internet
notification	soldier	bleed	download
selfie	troop	broken	wireless
account	army	scar	computer
timeline	enemy	hurting	email
follower	civilian	injury	virus

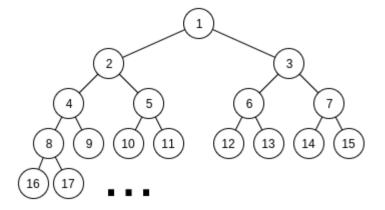


• For a given dependency tree *T*



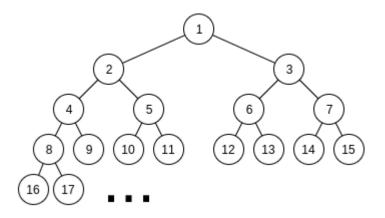


- For a given dependency tree *T*
 - Let *N* denotes set of nodes which are nouns / pronouns



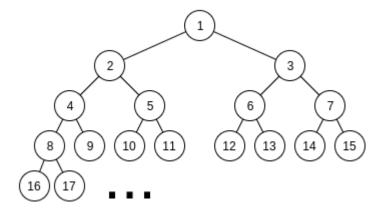


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 - Let E denotes set of nodes which belong to Empath dictionary



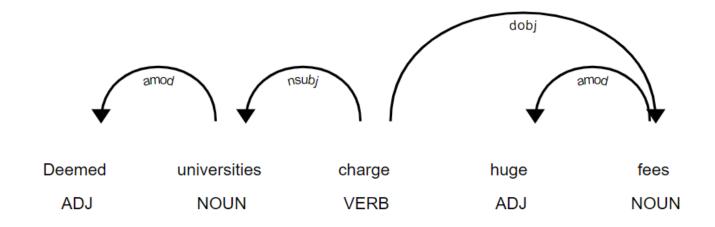


- For a given dependency tree *T*
 - Let *N* denotes set of nodes which are nouns / pronouns
 - Let E denotes set of nodes which belong to Empath dictionary
 - For $\forall n \in \mathbb{N}, e \in E$, if there exists a path $n \to e$, we call it a dependency path





Creating Dependency Paths



\mathcal{N}	\mathcal{E}	Dependency Path
fees	huge	$fees \xleftarrow{amod} huge$
universities	huge	$universities \xrightarrow{nsubj} charge \xleftarrow{dobj} fees \xleftarrow{amod} huge$



Identifying Most Discriminative Path for a Logical Fallacy

• From dependency tree, we created every feasible dependency path



Identifying Most Discriminative Path for a Logical Fallacy

- From dependency tree, we created every feasible dependency path
- Not all dependency paths are important



Identifying Most Discriminative Path for a Logical Fallacy

- From dependency tree, we created every feasible dependency path
- Not all dependency paths are important
- Need a way to select dependencies which are highly specific to a given fallacy



We take inspiration from KL divergence



We take inspiration from KL divergence

$$KL(P||Q) = \sum_{x} s(x) = \sum_{x} P(x) \log \left(\frac{P(x)}{Q(x)}\right)$$

Higher (more positive) the value of s(x), more discriminative x is for distribution P when compared to Q



ullet Let there be n logical fallacy classes, and x denote dependency paths



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• Then, score of path x in class C_i is given as

$$S_i(x) = \mathcal{P}_i(x) \log \left(\frac{\mathcal{P}_i(x)}{\mathcal{U}(x)}\right)$$



Fallacy of Relevance

Fallacy of	It is claimed that \mathcal{A} implies \mathcal{B} ,	
Relevance	whereas \mathcal{A} is unrelated to \mathcal{B} .	$1. \ nominal \ subject \rightarrow clausal \ comple-$ $ment \rightarrow nominal \ subject$
		$2. \ direct \ object \rightarrow nominal \ subject$
		3. direct object \rightarrow open clausal complement \rightarrow nominal subject
		4. object of preposition \rightarrow prepositional modifier \rightarrow clausal complement \rightarrow nominal subject
		5. nominal subject \rightarrow clausal complement \rightarrow clausal complement \rightarrow nominal subject



Ad Hominem

Ad	\mathcal{A} is claiming \mathcal{B} . \mathcal{A} is a moron. There-	
Hominem	fore, $\mathcal B$ is not true.	$1. \ \ adjectival \ \ modifier \ \rightarrow \ \ attribute \ \rightarrow \\ nominal \ subject$
		$2. \ \ attribute \rightarrow nominal \ subject$
		3. $adjectival\ complement\ o\ nominal$ $subject$
		$ \begin{array}{ll} 4. & adjectival & modifier \rightarrow direct & object \\ & \rightarrow nominal & subject \end{array} $
		5. $attribute \rightarrow clausal\ complement \rightarrow nominal\ subject$



Conclusion

• Logical fallacies refer to flawed patterns of reasoning



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- Fallacious users participating in Perspective debates are less when compared to For-against debates
 - They have more adversaries
 - They have more allies



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- Logical fallacies refer to flawed patterns of reasoning
- Fallacious users participating in Perspective debates are less when compared to For-against debates
 - They have more adversaries
 - They have more allies
- Proposed a novel approach for detecting logical fallacies



Future Works

 Our study proved that the linguistic structure of texts can be used as an alternative approach in identifying logical fallacy



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- These linguistic patterns can be used to create a static set of rules for identifying logical fallacy
- Future works can use these patterns and embed them in classification models for logical fallacy detection



thank you.