



Network-aware Fake News mitigation on social media

Scientific Report #2

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Introduction

Task: We want to detect fake news in social media posts containing both text and image.

Classification: 0 - Fake / 1 - Real

Introduction













Maryland drive gets probation for Delaware crash that killed 5 NJ family members

New 'Natural Feeding' trend has parents puking on babies

Neuroscience Says Doing This 1 Thing Makes You Just as Happy as Eating 2,000 Chocolate Bars

My plane hit an orca right after takeoff

Bowl of mussels

I just thought that was sitting in the deli

True

Satire/Paroday

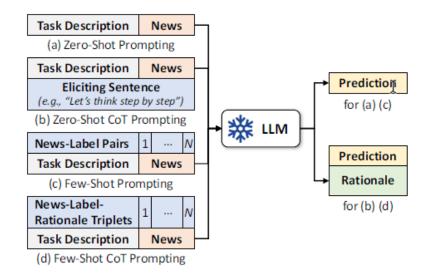
Misleading Content Manipulated Content

False Connection Imposter Content

Introduction

The main *objective* of this scientific report is to test the effectiveness of LLMs and prompt engineering tactics in the context of social media fake news detection, through experiments. We will also test the importance of adding visual information to the decision process.

Hu et al. [2] present their takes on the typical strategies like Zero-Shot and Few-Shot and the CoT [3] approach. They also introduce their hybrid model that passes the reasoning further to a trained BERT model, showing significant results.



Model	Usage	Chinese	English
GPT-3.5-turbo	Zero-Shot CoT	0.677	0.666
	from Perspective TD	0.667	0.611
	from Perspective CS	0.678	0.698
BERT	Fine-tuning	0.753	0.765
Ensemble	Majority Voting	0.735	0.724
	Oracle Voting	0.908	0.878

Figure 2: Results for article [2]

Zhang et al. [4] explored a step-by-step type of prompting in order to separate a claim into several subclaims and then verify them via progressive question answering. One of the fundamental problems of LLMs is hallucination, and the authors tackle this issue by providing real-time contextual information.

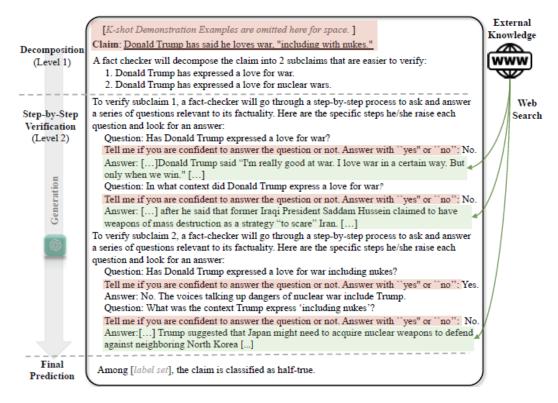
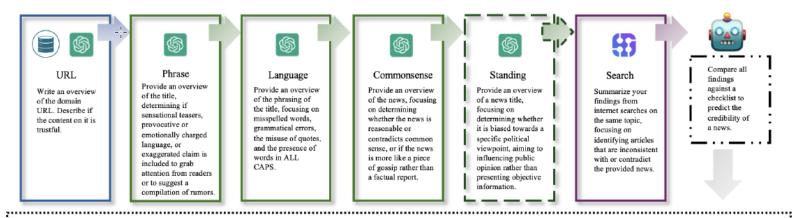


Figure 3: Step-by Step pipeline [4]

Li et al. [5] published another collection of experiments similar to the ones documented in [3]. Their contribution lies in a new way of gathering contextual information by using specific tools and leveraging the step-by-step method to extract the final answer.



Here is the checklist for fake news detection:

- 1. If the news is from a domain URL that is not well-known or requires skepticism, the news might be fake.
- 2. If the news language in the news title includes sensational teasers, provocative or emotionally charged language, or exaggerated claim to grab readers' attention or to suggest a compilation of rumors, it might be fake.
- 3. If the news title includes misspelled words, grammatical errors, the misuse of quotes, or the presence of words in ALL CAPS, it might be fake.
- 4. If the news is potentially unreasonable or contradicts common sense, or if the news is more like a piece of gossip rather than a factual report, it might be fake.
- 5. If the news is biased towards a specific political viewpoint, aiming to influence public opinion rather than presenting objective information, it might be fake.
- 6. If other online sources contain any inconsistent, conflicting or contradictory content, the news might be fake.

Here are the collected findings: {Observations}

Check each bullet point in the checklist above step by step to predict the veracity of the news {News}.

If the news is more likely to be a real news, return 'real';

otherwise, return 'fake'. Please refrain from providing ambiguous assessments such as undetermined.

Figure 4: Agent method pipeline [5]

This article [6], presents a unique way of leveraging the image context and up-to date web searching for an accurate prediction of fake news in the domain of politics.

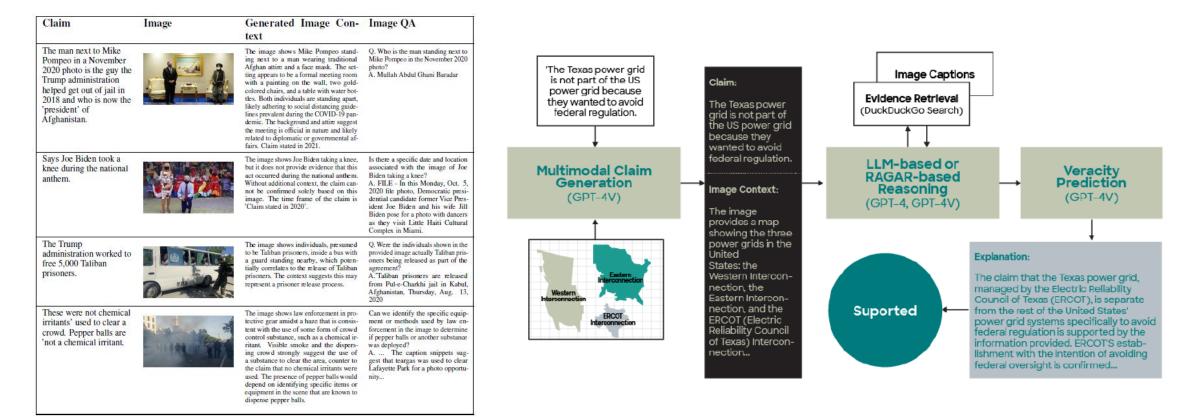


Figure 5: Example with claims, images and QA [6]

Figure 6: Detailed overview of the pipeline [6]

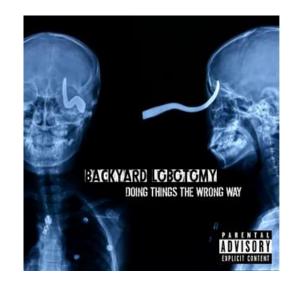
Datasets

The dataset used for testing was Fakeddit [1], along with Twitter15 and Twitter16 [13] for baselines. Fakeddit is a multimodal dataset containing over 1 million samples and 2-way, 3-way or 6-way classification labels.

More examples:



At my local community center they make disabled people get out the car to move a cone before they can park



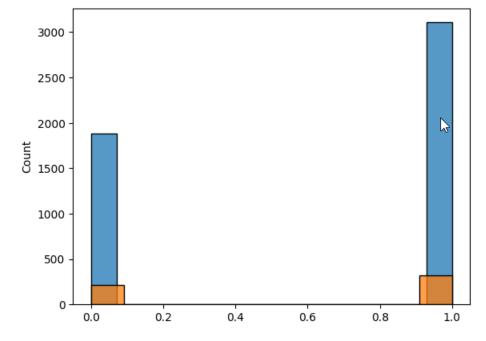
Backyard Lobotomy - Doing Things the Wrong Way

Datasets

Dataset Statistics	
Total samples	1,063,106
Fake samples	628,501
True samples	527,049
Multimodal samples	682,996
Subreddits	22
Unique users	358,504
Unique domains	24,203
Timespan	3/19/2008 - 10/24/2019
Mean words per submission	8.27
Mean comments per submission	17.94
Vocabulary size	175,566
Training set size	878,218
Validation set size	92,444
Released test set size	92,444
Unreleased set size	92,444

Dataset	Articles	Fake news	Real news	Unverified
Fakeddit	5633	2214	3419	0
Twitter15	1340	332	670	335
Twitter16	740	187	370	181

Used datasets statistics



Label distribution (2-way labels)

We propose a Few-Shot Agent Prompting method that leverages tools similar to the ones proposed by Li et al. [5] and the power of example that comes with few-shot prompting. The final assessment will be simple role-playing prompt, not using the step-by-step checklist presented in [5], due to its weak performance on our specific task.

Tools:

- 1. **Title tool**: a tool designed to determine if a social media post's title contains sensational teasers, provocative or emotionally charged language, or exaggerated claims to grab attention from readers or to suggest a compilation of rumors.
- 2. **Author tool**: a tool that provides insight into the author's username and the type of content posted on the respective subreddit that contains the post (can it be trustful?).
- 3. **Image tool**: a tool that analyzes a given image, trying to determine if it presents signs of alteration or meme-like characteristics or if it describes real events.

Title tool

Provide an overview of this social media post's title, determining if sensational teasers, provocative or emotionally charged language, or exaggerated claim is included to grab attention from readers or to suggest a compilation of

rumors.

Author tool

Provide an overview of this social media post's author username and the subreddit it has been posted to and determine if the content posted by the use in this thread could be trustful or not.

Image tool

Provide an overview of this social media post's image, determining wether it can describe real events or if it is edited, manipulated or meme-like rather than a real photograph.

Final assessment

Assume the role of a diligent judge in a fake news detection exercise, tasked with deciding if a post on the Reddit platform could potentially be fake news or not.

You will receive a post from the Reddit platform along with another LLM's reasoning.

If the reasoning matches the other given information, then respond with the respective 0 for fake news or satire or 1 for real news.

Remember to only answer with either 0 or 1 and no other reasoning.

Proposed pipeline



Examples

Title: EU is convinced it has found new way for young people to fall in love with neighboring countries free train rides...kicked off DiscoverEU project Thursday to send up to 30,000 18-year-olds chugging across continent this year, giving them free rail tickets to broaden their horizons.

Author: madazzahatAter Domain: abcnews.go.com Number of Comments: 7.0

Score: 75

Subreddit: upliftingnews

Upvote ratio: 0.89



The post is from a reputable domain (abcnews.go.com) and is posted in the subreddit upliftingnews, which typically shares positive and uplifting stories. The title also seems plausible and in line with positive news stories. The high upvote ratio of 0.89 and a score of 75 indicate that the post is well-received by the community.

Title: Robert Mueller resigns from the

Department of Justice (2019)

Author: RayBrower
Domain: i.imgur.com
Number of Comments: 0.0

Score: 30

Subreddit: fakehistoryporn

Upvote ratio: 0.9



The post is from the subreddit 'fakehistoryporn' which suggests that the content is likely satire or parody. Additionally, the title states that Robert Mueller resigned from the Department of Justice in 2019, which is not factually accurate. The high upvote ratio could indicate that users found the post entertaining rather than informative.

Tools output

(Label: Fake)

Title: We've got a new radiator

Author: Justrolledinshop_SS

Domain: imgur.com

Number of Comments: 20.0

Score: 100

Subreddit: subredditsimulator

Upvote ratio: 0.97

Title tool: The title \"We've got a new radiator\" appears to be straightforward and lacks sensational teasers, provocative language, or exaggerated claims. It does not suggest a compilation of rumors; instead, it seems to convey a simple update or announcement about a new radiator. The title is likely intended to inform rather than to provoke or sensationalize, making it more neutral in tone. Overall, it does not employ tactics commonly used to grab attention in a sensational manner.

Author tool: The author of the post, \"Justrolledinshop_SS,\" appears to be associated with a username that suggests a connection to a shop or business, possibly related to automotive or repair services, as \"Just Rolled In\" is a common phrase used in that context. However, without specific content from the post, it's difficult to assess the credibility of the author. The subreddit \"subredditsimulator\" is a unique community that simulates the experience of browsing various subreddits. Posts in this subreddit may not always reflect genuine content or discussions, as the nature of the subreddit is to create a simulated environment rather than a platform for serious discourse. Given these factors, the trustworthiness of the content posted by the user in this thread would likely be low. The context of the subreddit suggests that the posts may be more for entertainment or simulation rather than factual or reliable information. To make a more accurate assessment, it would be necessary to review the specific content of the post.

Image tool: The image depicts a car engine compartment with a visible battery and some tools or items nearby. The battery appears to be a Duralast brand, which is a common automotive battery. The overall condition of the engine bay looks somewhat dirty or neglected, suggesting it may be an older vehicle or one that hasn't been maintained recently. Based on the description, this image likely represents a real event, possibly related to car maintenance or repair. There are no obvious signs of manipulation or editing that would suggest it is a meme or altered photograph. However, without further context, it's difficult to determine the specific circumstances surrounding the image.

Final result: 0

Results

Model	Accuracy	Precision	Recall	F1
DistilBERT BiGruCNN	76.73	76.66	75.11	
Fine-tuned ViT	74.32	73.77	73.15	
Multimodal	77.70	77.27	76.72	
Few-Shot Agent	85.68			

Comparison with supervised learning models

Results

Model	Accuracy	Precision	Recall	F1-Score
Zero-Shot Prompting (GPT3.5)	76.52	0	0	0
Zero-Shot CoT Prompting (GPT3.5)	78.05	0	0	0
Few-Shot Prompting (GPT3.5)	67.17	0	0	0
Few-Shot CoT Prompting 4ex (GPT3.5)	78.96	0	0	0
Few-Shot CoT Prompting 8ex (GPT3.5)	83.74	0	0	0
Zero-Shot Tools RAG (GPT4mini)	70.71	0	0	0
Few-Shot Tools Prompting 8ex (GPT4mini)	85.68	0	0	0

Comparison between our LLM experiments

Results

Model	Accuracy	Precision	Recall	F1
MCWDST [4]	76.90	77.27	76.89	76.82
Fakeddit [8]	89.09	-	-	-
DEAP-FAKED [3]	89.55	-	-	-
(BERT+Dense)+ Xception [9]	91.87	93.39	93.29	93.25
Multimodal transformers [10]	92.51	93.83	93.74	93.79
LLM+SLM [3]	78.60	78.40	81.40	80.40
FactAgent with Expert Workflow [5]	88.00	88.00	89.00	88.00
ToRAG + CoTVP + CoVe [6]	84.00	85.00	86.00	85.50
Ours	85.68	-	-	-

Comparison with state-of-the-art

Conclusion

In conclusion, we confirmed that large language models have a substantial capability to detect fake news. The multimodal aspect of the task seemed to help the model achieve better reasoning and comparing the context of the image with the news title can be a future area to improve on.

The main upside of working with LLMs is that they don't need training and can run at a relatively low cost. As a solution for a social media company that can either have its own LLM or a subscription to one already, it proves to be essential to take into consideration when mentioning the automatic fake news detection task.

Of course, all of this does not come without its flaws. The somewhat randomness of the model's answers and the web searches are the main points of concern. Also, we need to take into consideration the hallucination problem.

Further experiments can improve this kind of detection and we will continue with it.

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