# War of Words: Using Large Language Models and Retrieval Augmented Generation to Classify, Counter and Diffuse Hate Speech

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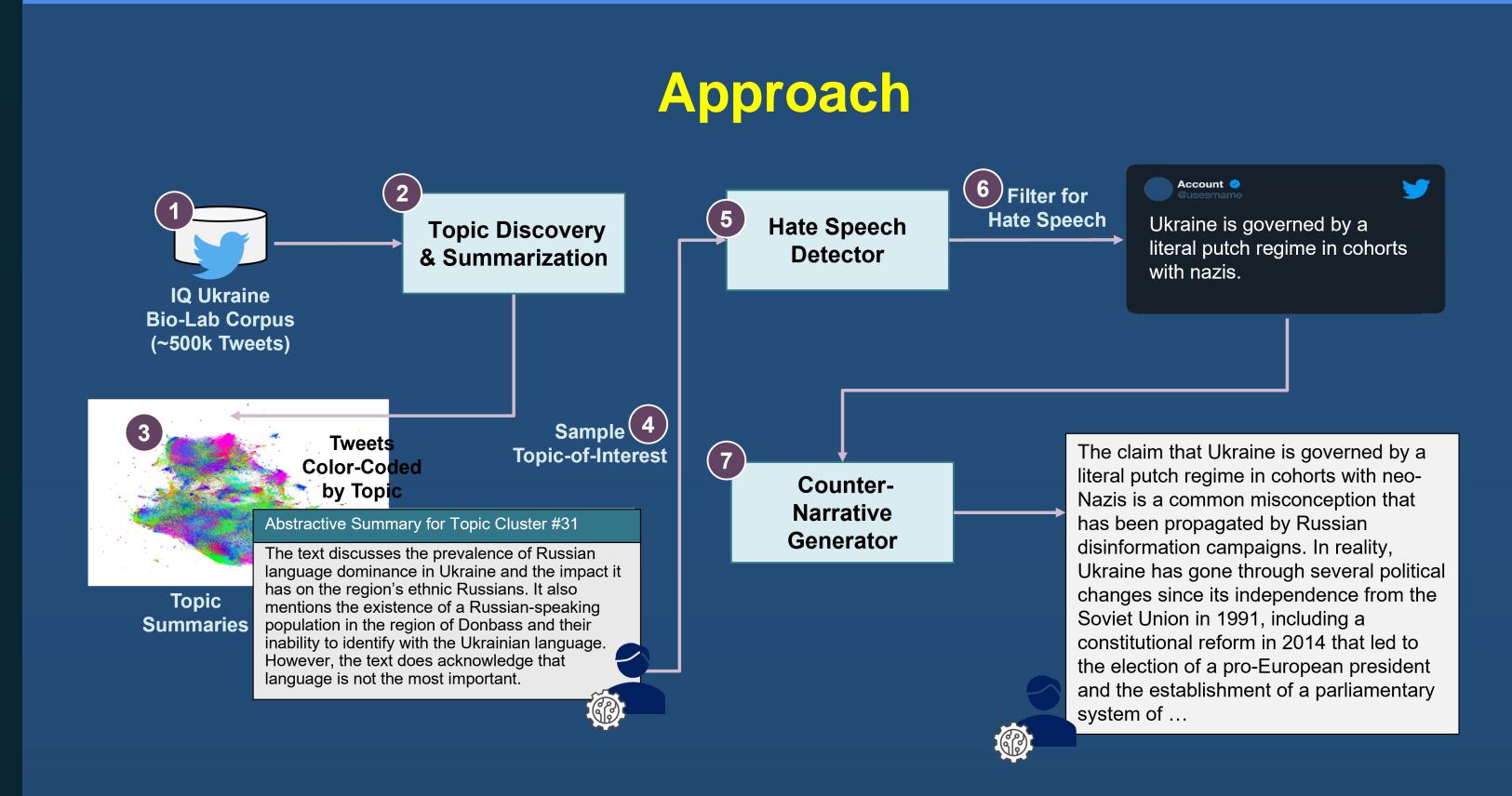
# Introduction

#### MOTIVATION

There is a need for a robust and effective defense strategy against malign influence spread by influence operation campaigns perpetrated by foreign state actors and other entities

## PROBLEM

- Social media and, more recently, generative AI make it very easy to generate disinformation and hateful content at scale
- Monitoring via traditional topic/narrative modeling often focuses on low-level content that can be difficult to interpret
- Highly manual workflows typically used for content prioritization and response generation do not scale



Counter-Narrative Generation Prototype

# CLASSIFICATION OF SOCIAL MEDIA POSTS CONTAINING INFLAMATORY SPEECH

- Identify extreme content via hate-speech classifier and rank-order tweets in topic clusters
- Zero-shot prompt-based prediction using Mistral Instruct & Twitter's hate speech guidelines
- ~1k hand-labeled evaluation set
- Compare to pre-trained HateBERT1, Fine-Tuned RoBERTa2, prompted LLaMA-7b/LLaMA2-7b



Hate Speech Classification Pipeline

# AUTOMATED RESPONSE TO INFLAMATORY SPEECH

- Retrieval Augmented Generation (RAG) using relevant factual context (Wikipedia, News Articles) to ground auto-generated response (via FAISS over MPNet paragraph embeddings)
- Mistral-7B prompted to make use of factual arguments utilizing the RAG database to generate contextually relevant and correct counter-speech
- Human evaluation on Likert scale of 5 generations for each of 20 random hate speech tweets; 2 raters, 5 quality dimensions



Counter-Narrative Generation Pipeline

## Results

### DATASET AND METRICS FOR EVALUATING HATE SPEECH DETECTION

- Scraped tweets related to Ukraine war and bio-weapons labs during a period leading up to the war
- Time period: between December 2021 and January 2022
- After filtering and removing duplicates, dataset contained ~500k unique tweets

Model	Precision ( <b>û</b> )	Recall (①)	F1 score (①)	Accuracy (압)	Time to run in mins (⇩)
HateBERT Pre-Trained	0.00	0.00	0.00	0.63	117
RoBERTa Fine-Tuned	0.84	0.35	0.49	0.73	105
LLaMA-7B Zero-Shot	0.38	1.00	0.54	0.38	240
LLaMA2-7B Zero-Shot	0.90	0.96	0.93	0.95	102
Mistral-Instruct Prompted	0.96	0.97	0.97	0.97	28

Hate Speech Detection Evaluation (BioLab Eval Set)

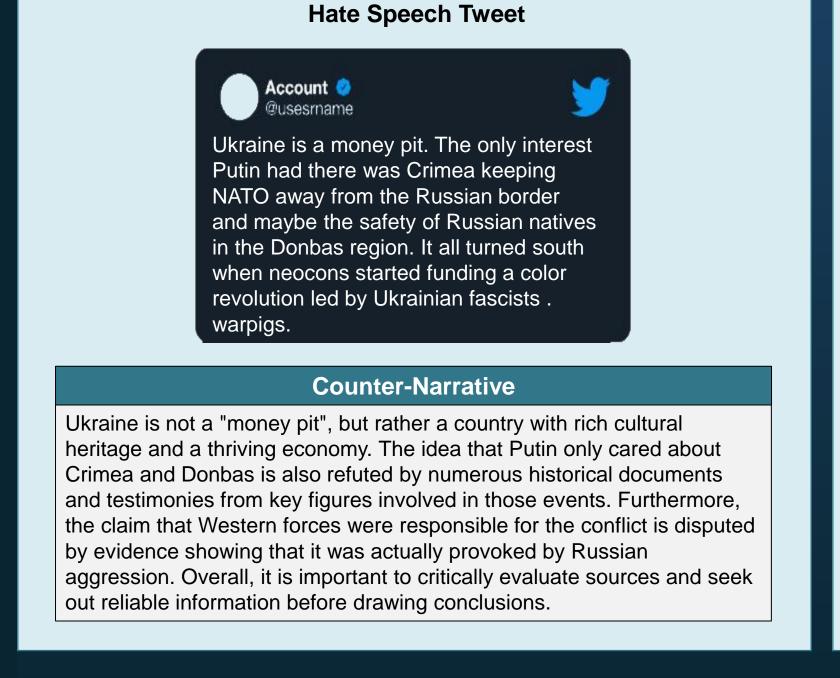
### METRICS FOR EVALUATING COUNTER-SPEECH GENERATION

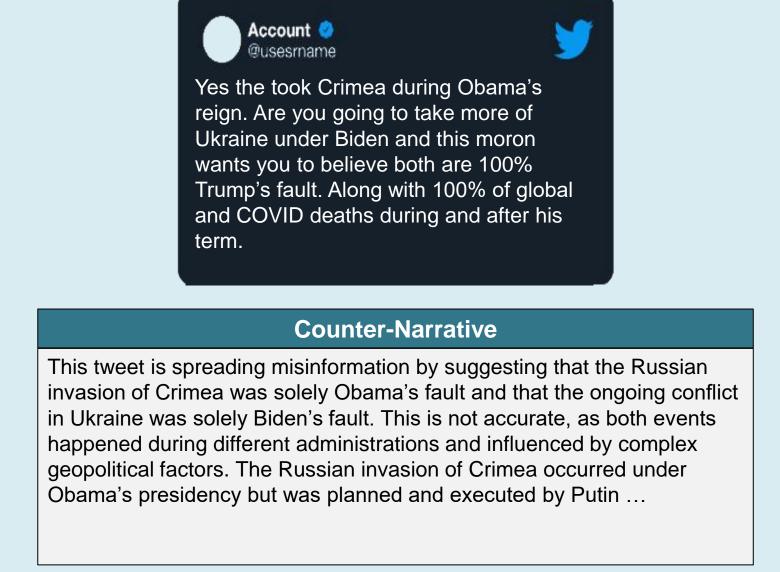
- Produced five unique counter-narrative samples for each of 20 randomly selected hateful tweets, resulting in a total of 100 counter-speech samples
- Manually evaluated each counter narrative along 4 dimensions: factuality, relevance, grammaticality and diversity using 1(bad) to 5(good) scale
- Note: one diversity score was assigned for all five counter-narratives responding to a hate tweet
- To ensure an unbiased assessment, two independent raters evaluated the same 100 counter- speech samples
- Inter-rater reliability was quantified using Cohen's Kappa statistic

Metric	Mean (①)	Median (압)	Cohen Kappa (①)
Factuality	3.6	4	o.68
Relevance	3.8	5	0.76
Grammaticality	4.4	5	0.80
Diversity	3.7	5	0.79

Counter-Speech Evaluation Metrics

# EXAMPLES OF AUTOMATICALLY GENERATED COUNTER-NARRATIVES





**Hate Speech Tweet** 

# Conclusion

- Our research demonstrates the effectiveness of zero-shot learning and LLM prompt engineering for nuanced hate speech classification surpassing prior state-of-the-art models
- Our approach effectively combines RAG's information retrieval with LLMs' context processing, overcoming the biases of traditional models and excels in generating coherent and to a large extent relevant and factual counter-narratives

# FUTURE WORK

- Reduce LLM-generated hallucinations (i.e., untruthful information and claims) by expanding RAG database and other approaches
- Develop effective techniques for prompt engineering in order to improve relevance of the generated counter-narratives
- Investigate Langchain's capability to incorporate chat history iterations in order to generate conversational response