



Hidden Persuasion: Detecting Manipulative Narratives on Social Media During the 2022 Russian Invasion of Ukraine

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73% of Ukrainians use Telegram as a primary source of information (USAID-Internews, summer, 2024):



Are we reading facts or interpretations disguised as facts?

путін грає з вогнем: він не розуміє, що якби не я, з росією вже сталося б багато дуже поганих речей, і я маю на увазі ДУЖЕ поганих, – Трамп. Риторика Президента США змінюється останніми днями, побачимо, чи зміниться

Як це вплине на ситуацію в Україні?



щось у діях



🤨 1,9K

●● 487



409

236

209

83

● 831,4K edited 11:45

Are we reading facts or interpretations disguised as facts?



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Are we reading facts or interpretations disguised as facts?



1 million people viewed the post

Problem formulation

Goal:

Detect manipulative narratives in Ukrainian Telegram posts.

Manipulation defined (UNLP 2025):

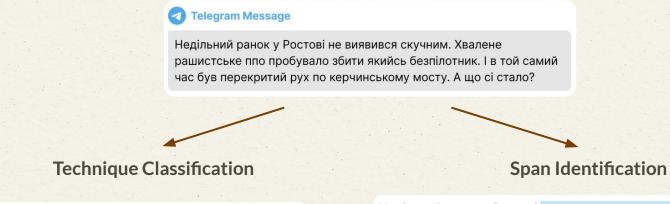
The **use of rhetorical or stylistic techniques** to influence readers' opinions or behavior — **without relying on factual evidence**.

Problem formulation

euphoria

loaded language

Tasks



cliche

Недільний ранок у Ростові **не виявився скучним. Хвалене** рашистське ппо пробувало збити якийсь безпілотник. І в той самий час був перекритий рух по керчинському мосту. А **що сі стало**?

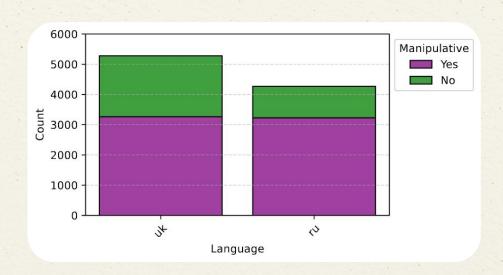
Data Overview

- Source: Telegram posts from Ukrainian channels (provided by UNLP 2025 Shared Task)
- Each post is annotated with manipulation techniques and manipulative spans

• 2 languages present:

Ukrainian: 5,278 posts

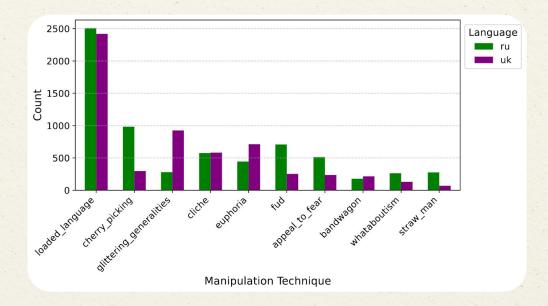
Russian: 4,269 posts



Data Overview

- Source: Telegram posts from Ukrainian channels (provided by UNLP 2025 Shared Task)
- Each post is annotated with manipulation techniques and manipulative spans

• 10 predefined techniques:



Proposed Solution - Technique Classification

Fine-tuning with Causal Language Modeling

Fine-tuning for **Multilabel** Sequence **Classification**

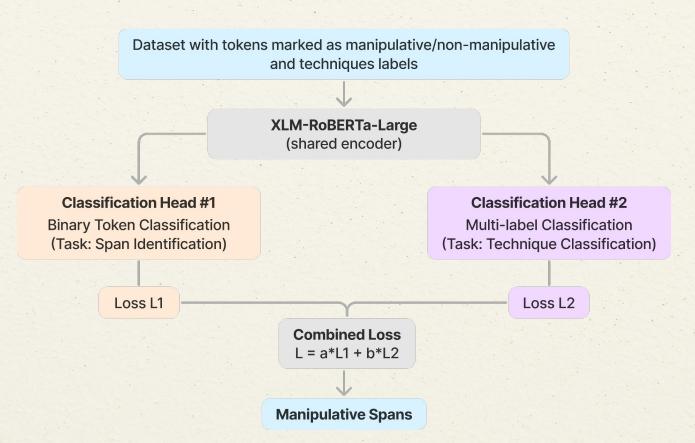
CatBoost Post-processing

Trained LoRA adapter

Class probabilities

Predictions

Proposed Solution for Span Identification



Results

Metrics

Technique Classification

macro-averaged F1 score

Span Identification

span-level F1-score

Results - Technique Classification

Team	Public	Private
GA	0.47369	0.49439
MolodiAmbitni (Gemma2 w/PP)	0.46203	0.46952
MolodiAmbitni (Gemma2)	0.43933	0.45543
CVisBetter_SEU	0.43669	0.45519

Comparison of metrics for top-3 solutions from competition leaderboard

Results - Technique Classification

Technique	F1 score	Support
loaded_language	0.782	2959
glittering_generalities	0.644	723
euphoria	0.550	695
fud	0.525	576
cherry_picking	0.467	768
appeal_to_fear	0.450	449
cliche	0.328	695
whataboutism	0.296	235
straw_man	0.287	207
bandwagon	0.215	236

Results - Span Identification

Solution	Public	Private
GA	0.64598	0.64058
CVisBetter_SEU	0.59873	0.60456
MolodiAmbitni	0.59662	0.60001
OpenBabylon	0.59142	0.59096
MolodiAmbitni (baseline)	0.58617	0.58794

Comparison of metrics for top-4 solutions from competition leaderboard

Conclusions

- Achieved 2nd place in technique classification and 3rd place in span detection in the UNLP 2025 Shared Task
- Developed fine-tuned Gemma 2 + meta-feature post-processing that significantly boosted classification performance
- Showed that a simple XLM-RoBERTa model, paired with a dual-head pipeline, can achieve top-tier span detection results

Thank you! Q&A

Stage 1: Fine-tuning with CLM

Model

Gemma 2B IT



Training Setup

LoRA (Alpha (α): 32, Rank (r): 32) + 4-bit quantization

in causal LM setup

Output

LoRA adapter for techniques generation

Stage 1: Fine-tuning with CLM

Prompt Composition

System:

You are an AI trained to detect rhetorical manipulation in social media. Return ONLY the technique names from the list, comma-separated.

User:

Task: Identify techniques in this post using ONLY the following: <techniques description>

Examples: <2 examples of posts + their techniques>

POST to analyze: <target post text>

Assistant:

Predicted output: <technique1, technique2, ...>

Stage 2: Supervised Multi-label Classification

Model: Gemma 2B IT + LoRA adapter from Stage 1

Training Setup:

- 1. Multi-label sequence classification
- 2. LoRA (Alpha (α): 32, Rank (r): 16) adapter
- 3. Threshold selection per class

Output: class probabilities for each text

Stage 3: Post-Processing

Model Used: CatBoost

Features Used:

- technique probabilities from Stage 2
- cosine distances -> current text and centroids of trigger phrase clusters
- **frequency** of techniques among top-20 **nearest texts** and **trigger phrases**

meta-features:
 word count, number of question marks, presence of URLs, etc.