

Unveiling the dynamics of media discourse on NATO in Romania amid electoral turmoil through semantic network analysis

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Background & Research Questions

Romania's annulled presidential election of 24 Nov 2024, won by a populist outsider amid allegations of Russian interference, propelled NATO into the national spotlight. Network Agenda Setting (NAS) reframes classic agenda-setting by modelling media attributes as networks that shape public and policy agendas¹⁻³. In Eastern Europe, NAS studies remain scarce⁴ while most regional work analyzes political discourse on Twitter⁵⁻⁶, overlooking mainstream news and Facebook, Romanians' main social media platform for news⁷. Integrating NAS with semantic network analysis⁸⁻¹¹ lets us test how Romania's online news agenda on NATO shifted during the 30-day pre- and post-election window.

Research Questions

1. **RQ1:** How does the semantic structure of NATO discourse differ 30 days *before* versus *after* the election?
2. **RQ2:** Which actors and entity clusters dominated each period?
3. **RQ3:** To what extent does anti-NATO discourse increase post-election?

Data & Methods

CORPUS. Romanian-language news articles and public Facebook posts containing “NATO” harvested **24 Oct – 24 Nov** (pre-election) and **25 Nov – 25 Dec 2024** (post-election).

- Pre-election: **3,234** mentions (News = 1,479, FB = 1,755)
- Post-election: **6,662** mentions (News = 3,068, FB = 3,594)

MONITORING. Data collected with NewsVibe¹² (v 25.1.14) (Social Monitor, 2025). Export snapshot: 30 Jan 2025.

IN-PLATFORM NLP PREPROCESSING. NewsVibe applies a fine-tuned NER model based on Llama3.2-1B. The model was trained on a synthetic dataset of news articles from the platform automatically annotated with Llama3.3-70b¹³ with the following entity types: Person, Location, Institution, Company, and Product. The fine-tuned model obtained an F1 score of 82.5%.

NETWORK ANALYSIS. The graph constructed using the list of entities from each article is imported into Gephi^{14,15} (v.0.10).

- Nodes = entities; edges = frequency-weighted co-occurrences

- We analyse: Graph size, Modularity¹⁶, Top-10 rankings by degree and weighted degree

STANCE PROXY MARKERS. Anti-NATO salience is gauged using two markers: (1) the total occurrences of *Călin Georgescu*, the anti-NATO candidate, and (2) the average weighted degree of the three Romanian military base nodes (*Deveselu*, *Cincu*, *Kogălniceanu*). Higher values of these markers can reflect greater anti-NATO emphasis in the discourse.

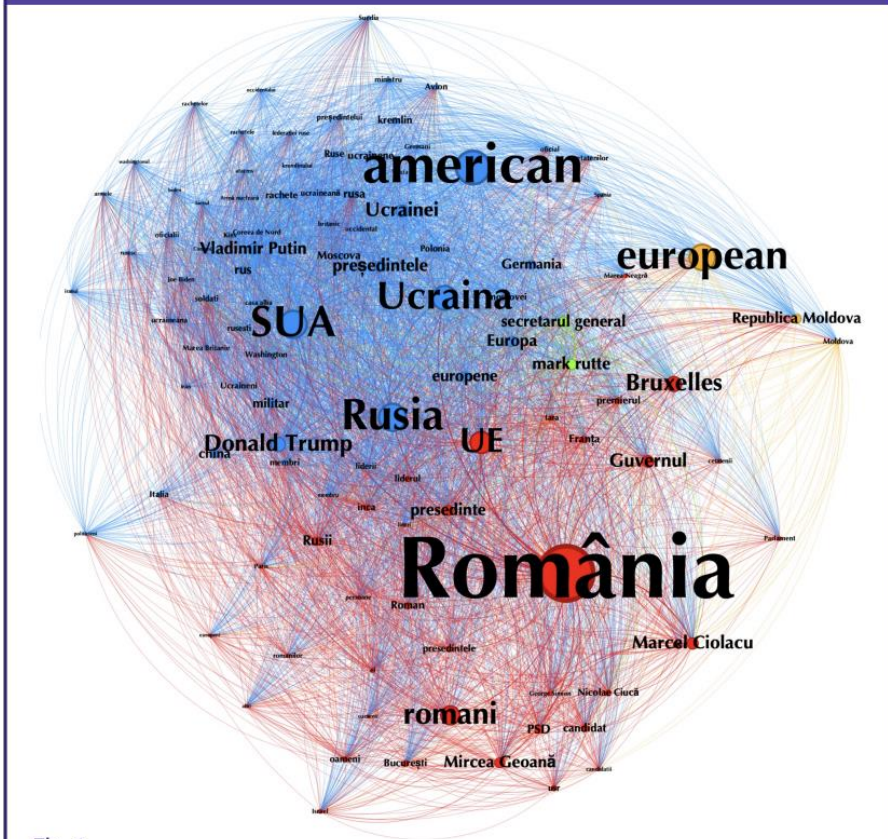
Results

METRIC	BEFORE THE ELECTIONS (Fig. 1)	AFTER THE ELECTIONS (Fig. 2)
Density	0.142	0.128
Louvain ($\gamma = 1$)	Q = 0.29 (5 communities)	Q = 0.20 (7 communities)
Size	Vertices = 730; Edges = 37,564	Vertices = 1,218; Edges = 94,161
Top-10 Degree	Romania (599); USA (543); Russia (540); Ukraine (536); EU (487); American (486); D. Trump (439); Europeans (428); Europe (417); Ukrainian (406)	Romania (1,104); Russia (1,052); EU (1,005); Ukraine (953); USA (920); Europeans (897); C. Georgescu (896); Romanians (844); European (809)
Top-10 Weighted Degree	Romania (9,739); Russia (7,614); Ukraine (7,424); USA (6,927); D. Trump (5,581); V. Putin (5,119); Ukrainian (3,929); EU (3,903); American (3,830); president (3,256)	Romania (35,145); Russia (24,055); EU (22,010); Ukraine (19,686); USA (15,443); C. Georgescu (11,691); V. Putin (11,151); D. Trump (10,315); Europeans (8,772); Europe (8,699)

KEY CLUSTERS

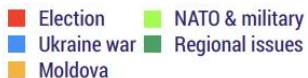
BEFORE THE ELECTIONS (Fig. 1)	AFTER THE ELECTIONS (Fig. 2)
Election (327 nodes, 44.79%); (general references to candidates' discourse)	Election (632 nodes, 51.89%); (impact of C. Georgescu on NATO relationship)
Ukraine war (265 nodes, 36.3%); (general info)	Ukraine war (410 nodes, 33.66%); (focus on C. Georgescu messages – opposition to Ukraine aid; alleges NATO blocked him to prolong war & involve Romania)
Moldova (74 nodes, 10.14%); (Russia accuses NATO of using Moldova in Ukraine war; electoral messages)	NATO & military 1 (102 nodes, 8.37%); (references to activities, exercises etc)
NATO & military (52 nodes, 7.12%); (references to activities, exercises etc)	NATO & military 2 (53 nodes, 4.35%); (references to activities, exercises etc)
Regional issues (12 nodes, 1.64%)	Russia (10 nodes, 0.82%); (allegations of interference in elections; regional threat)
	China (6 nodes, 0.49%); (allegations – TikTok helped influence elections in NATO country)
	Romanian Academy (5 nodes, 0.41%); (reacts to accusations of anti-NATO support)

BEFORE the elections

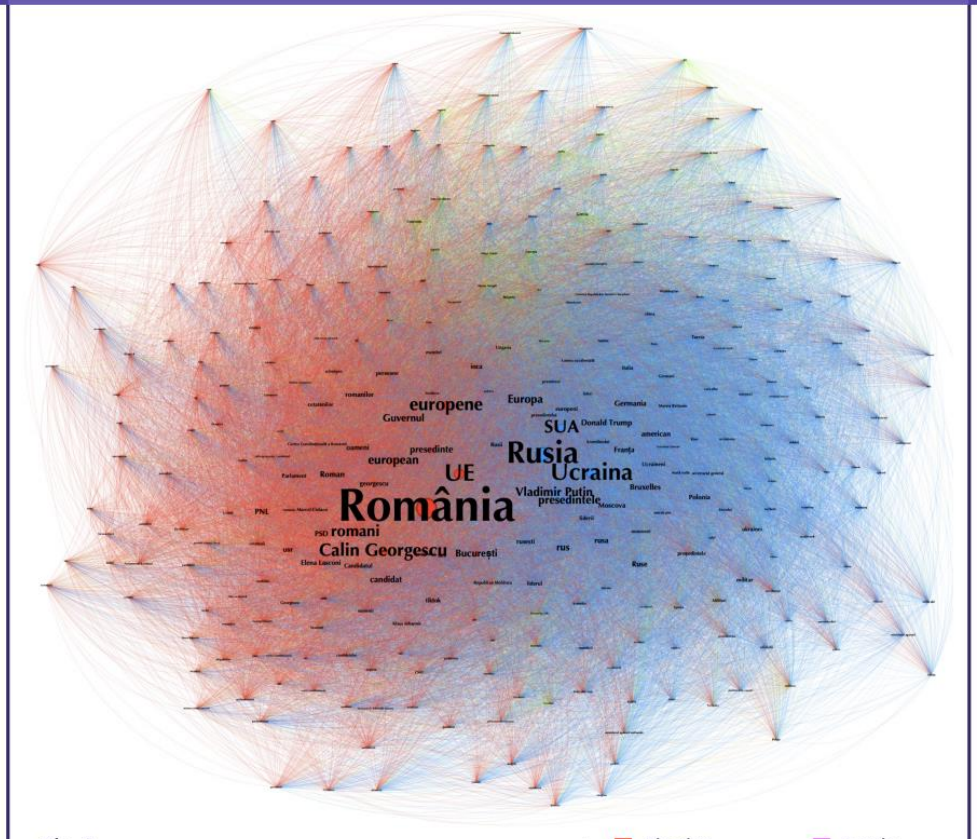


Semantic network (24 Oct 2024 – 24 Nov 2024).

Only nodes with degree > 200 (top 15%, n = 107) are displayed.
Node & label size proportional to degree.

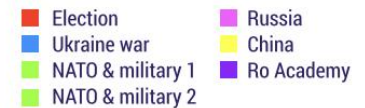


AFTER the elections



Semantic network (25 Nov 2024 – 25 Dec 2024).

Only nodes with degree > 200 (top 25%, n = 306) are displayed.
Node & label size proportional to degree.



Anti-NATO marker surge. The sharp rise of a single anti-NATO candidate and the amplified focus on local military installations signal a qualitative shift from diffuse criticism to concrete, actor-centred opposition.

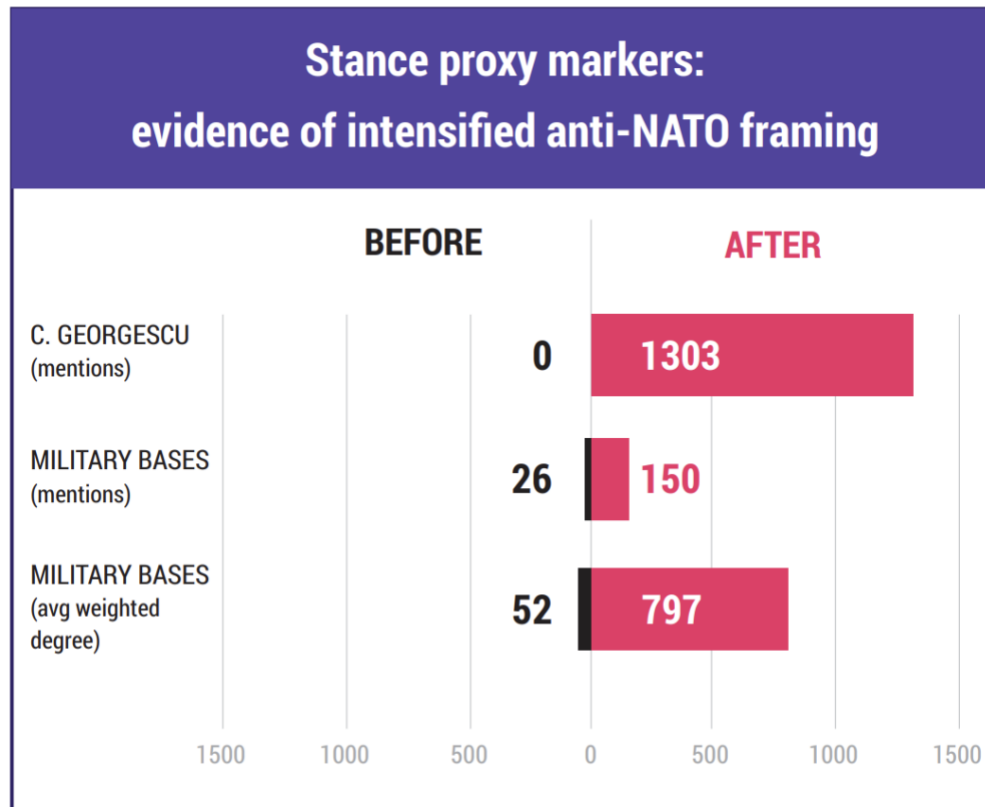


Figure 3. Stance proxy markers: evidence of intensified anti-NATO framing.

Key Findings

- NATO mentions doubled (3,234 → 6,662) post-election.
- Semantic network nearly doubled in size (730 → 1,218 nodes).
- Narrative clusters became less distinct: modularity drops ($Q = 0.29 \rightarrow 0.20$).
- Anti-NATO candidate emerged prominently (0 → 1,303 mentions, high degree node).
- Military-base prominence surged (52 → 797 avg. weighted degree).
- Post-election, the ELECTION cluster almost doubled to 632 nodes; the UKRAINE WAR expanded by 54% (major focus on Georgescu allegations that NATO seeks to drag Romania into war). *Russia* references 809 → 1,910 plus a standalone RUSSIA cluster (10 nodes). NATO & MILITARY clusters 46 → 146.

Conclusions

Actor realignment (RQ2): Post-election, anti-NATO voices and messages moved from the network periphery to its core

Structural & framing shift (RQ1 & RQ3): Network metrics show NATO discourse became denser (nodes ↑, edges ↑, modularity ↓); nodes associated with negative frames posted steep gains in either occurrence or weighted degree, signalling a stronger anti-NATO emphasis.

Methodological payoff: Integrating real-time monitoring with NewsVibe and semantic network analysis in Gephi proved effective for detecting these rapid, platform-spanning agenda shifts across Romanian news sites and Facebook.



- Total coverage of the tool: 7,500+ news sites, public FB pages, YouTube channels across the EU and the US.
- Study dataset: 5,000+ Romanian News websites + Romanian public Facebook pages
- Refresh: 10 min; NER: fine-tuned model based on Llama3.2-1B. Precision: 0.8315, Recall: 0.8192, F1: 0.8253

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DISCLAIMER

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Authors 1 & 3 are founder-architects of NewsVibe; Author 2 - NewsVibe software engineer; Author 4 - academic advisor.