

# Bridging the Narrative Divide: Cross-Platform Discourse Networks in Fragmented Ecosystems

*Narratives don't respect platform boundaries — our models shouldn't either.*

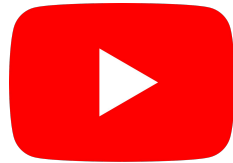


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# The Same Narrative, Told Across Many Platforms At Once

**January 6th, 2021:** "Trump said it's gonna be wild — pack your sh\*\*, we're heading to DC."



Source: NPR

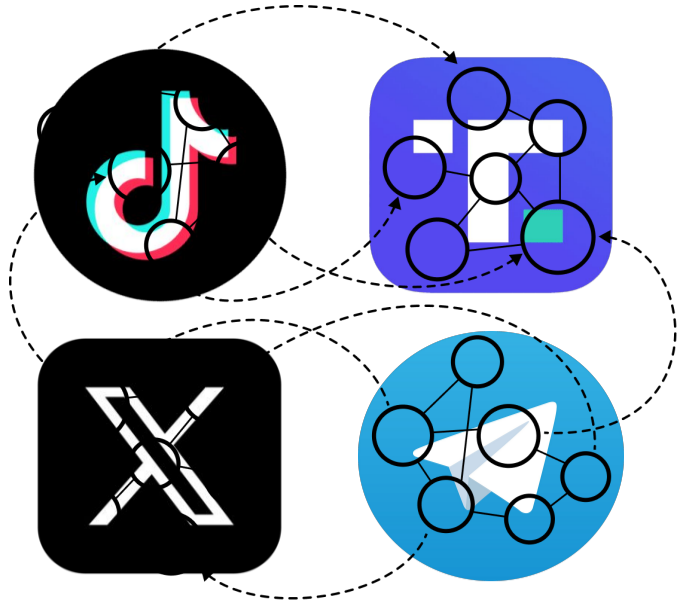
# The Same Narrative, Told Across Many Platforms At Once

**Springfield, OH, 2024:** “They’re eating the dogs, they’re eating the cats.” A local rumor that Haitian immigrants were eating pets started on Facebook.



Source: NPR

# Modern Narratives Take Shape *Across* Platforms



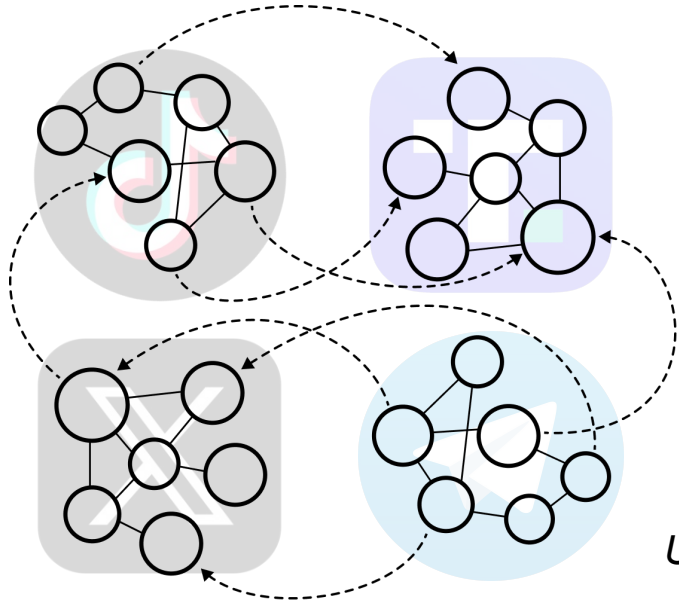
The online information ecosystem is fragmented.

- Users consume and share content across many platforms.
- The **borders between platforms are arbitrary — ideas and narratives do not respect them.**
- Without cross-platform visibility, we fail to anticipate real-world impacts.

# Why Traditional Methods Fail

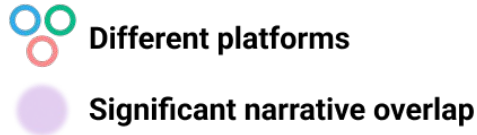
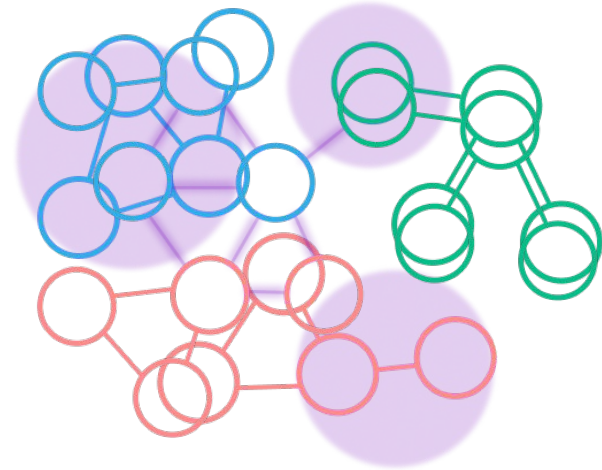
Traditional networks break at the platform boundary.

- Follower graphs, reposts, hashtags → **platform-locked**
- Semantic similarity → **brittle** under platform-specific language
- APIs → **disappearing**, incomplete, inconsistent



*Understanding modern information flow requires understanding how users connect across platforms, not just within them.*

# Rethinking User Representation Through Shared Discourse



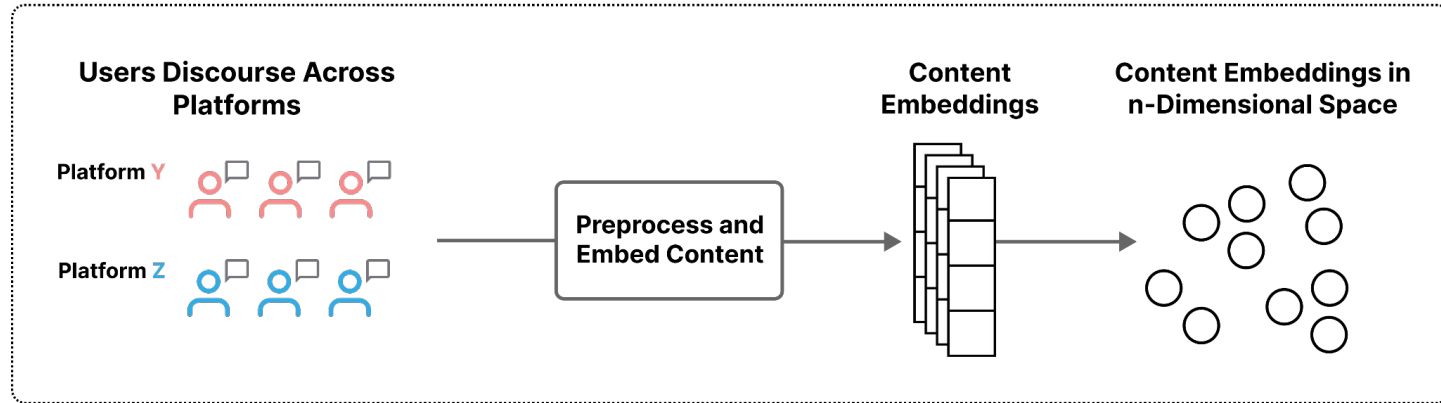
Instead of "who did this user repost?" → "what ideas does this user inhabit?"

- Each user = **distribution over latent narratives**
- Users who engage with the same narratives connect — regardless of platform
- No follower ties needed or shared metadata needed.

Content → Embeddings → Narratives → User Distribution → Unified Network

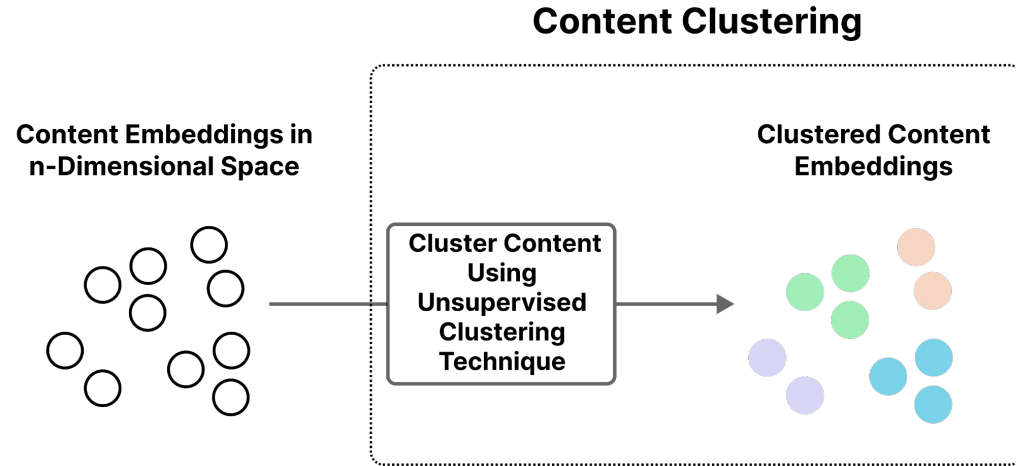
# Our Approach: Modeling Discourse as Network

## Content Aggregation and Processing



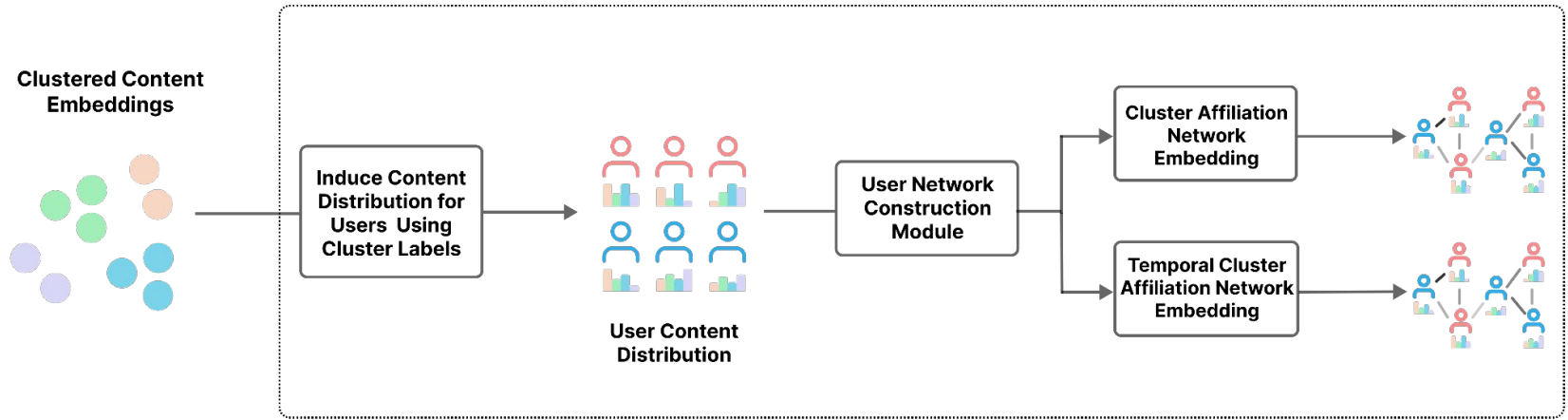
**Step 1: Content Aggregation and Processing** *Collect cross-platform discourse and transform text into comparable embeddings*

# Our Approach: Modeling Discourse as Network



**Step 2: Content Clustering** *Group similar discourse into thematic clusters using unsupervised learning*

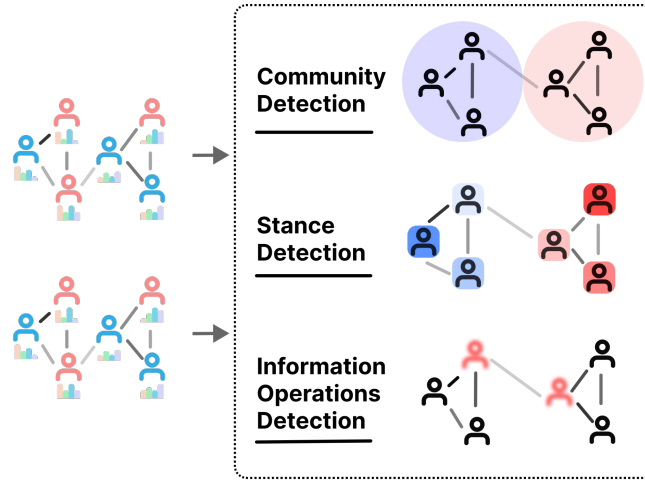
# Our Approach: Modeling Discourse as Network



## User Network Construction

**Step 3: User Network Construction** *Transform content clusters into user networks based on shared discourse patterns*

# Our Approach: Modeling Discourse as Network



## Downstream Tasks

**Step 4: Network Applications** *Use the constructed networks for traditional network analysis and detection tasks*

# Does it Work? Let's Test

We benchmark our approach against classic network construction methods on established and novel CSS tasks using a consistent set of classifiers.

Network Type	How Users Are Connected
<b>Co-Repost</b>	Repost the same content
<b>Co-URL</b>	Share the same URLs
<b>Fast Repost</b>	Repost identical content within a short window
<b>Hashtag Sequence</b>	Use the same ordered hashtag sequences
<b>Text Similarity</b>	Post highly similar text
<b>k-NN Embedding Graph</b>	Close in overall text embedding space
<b>Fused Graph</b>	Linked if connected in any of the above networks

## **Influence Operation Detection**

Detecting covert accounts that promote coordinated influence campaigns (Seckin et al. 2024).

## **Ideological Stance Prediction**

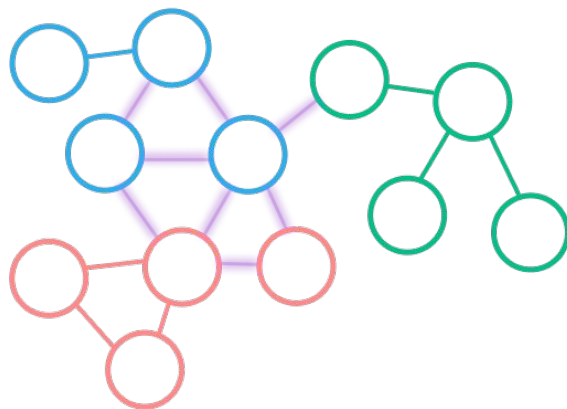
Inferring a user's political leaning based on their network position.

## **Cross-Platform Engagement Prediction**

Predicting which users will engage with content as it moves across platforms.

# Results: Leaner, broader, and consistently best-in-class

Discourse Networks perform better, need less data, and cover more users



Different platforms



Significant narrative overlap

## IO Detection

11% F1, 24% AUC improvement over baseline

## Ideological Classification

15-20% F1 improvement over best baseline

## Cross-Platform Engagement:

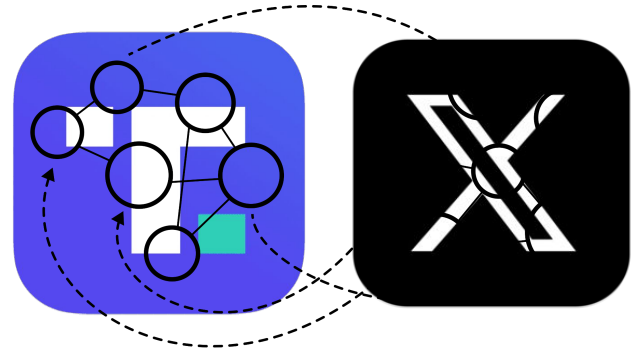
47% AUC improvement over best baseline

*Reaches peak performance with a fraction of users content; covers nearly all users.*

# The Discourse Network Exposes a Hidden Bridge Zone

X + Truth Social, 2024 U.S. Election: A Tiny Bridge Zone Spans Both Worlds

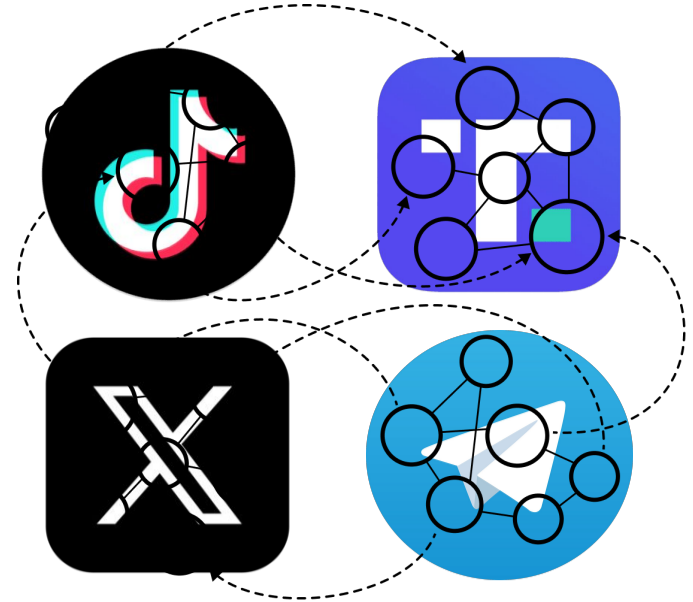
- One high-entropy community detected: 0.33% of users, 2.14% of posts
- First cross-platform carriers for 68–69% of migrating narratives
- Activity near platform median — signal comes from structural position, not volume
- Invisible to every other network tested



# Future Work: Those Bridges Are Predictive (WWW '26)

Discourse networks turn cross-platform diffusion into a proximity problem

- Extended to 5.7M posts across X, TikTok, Truth Social, and Telegram
- Users with active discourse neighbors are 22.4× more likely to adopt a narrative (OR, 95% CI [16.9, 29.7])
- Monitoring just 2.9% of users — those with high cross-platform connectivity — retains 97% of prediction performance



# Key Takeaways: Birds of a Feather *Think* Together

## Discourse-based graphs outperform traditional ones

Discourse-based networks capture **more users** with **less data**, while matching or exceeding traditional methods on standard tasks.

## Cross-platform by design

The approach **works across fragmented ecosystems**, without needing reposts, URLs, or user overlap.

## Bridge users in the wild

In the 2024 US election case study, a hidden **0.33 percent** of users **drove nearly 70 percent** of narrative migration between Truth Social and X.

```
pip install discourse-networks
```



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