Analyzing Public Discussions for Product Insights

Mining Reddit and Tiki for product issues, sentiment, and trends

From noisy threads \rightarrow actionable product signals.

[VISION]

Problem & Value

- Which products break, why, and how often—using public discussions.
- Turn unstructured Reddit and Tiki text into actionable product signals.
- Output insights teams can ship on: issues, trends, severity, examples.

Objectives

- Which issues? How frequent? How trending? What severity?
- Where do issues cluster (by product/subreddit/time)?
- How reliable are methods vs baselines?
- Deliver a repeatable pipeline and clear, prioritized findings.

Why These Platforms?

- Reddit: rich, threaded discussions; public API; strong NLP support.
- Tiki (e-commerce): structured, purchase-verified reviews; VN market signal.
- Complementary: text-rich threads vs short reviews \rightarrow broader coverage.

Data Source Comparison

Platform	Pros	Cons
Reddit	 Public API (easy access) Rich text discussions Strong support for English NLP 	 Some noise and off-topic posts Limited Vietnamese data API usage restrictions/quotas

	focused "subreddits"	
Facebook	 Large user base in Vietnam Active groups and communities Rich variety of topics 	 Many bots/spam accounts API restrictions Difficult to collect clean data Weak support for English NLP libraries
TikTok	Very popular among young usersStrong trend/meme insights	 Mostly media (videos, images) No official API Lacks group/community structure Scraping is slow (parse HTML)

• Hard to extract relevant textual data

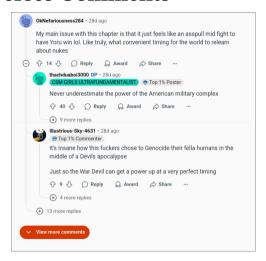
• Community-driven, topic-

Data Collection

- Initial: PRAW (Python Reddit API Wrapper) for prototyping.
 - Limitation: 1,000 post cap, rate limits.
- Current (Hybrid):
 - Reddit historical archives (Academic Torrents) to bypass API caps.
 - ► Tiki review dumps for cross-source validation.
 - **Result:** Broader time windows, more volume.

How PRAW Traverses Comments

- Reddit returns comment trees with View more comments placeholders.
- We expand lazily, traverse, and serialize
 —no one-click "download".
- Captures full depth where needed; avoids rate-limit explosions.



Alternatives Considered

Pushshift.io (Pushshift API)

- More powerful than PRAW, can filter posts by time
- Requires moderator status
 - \rightarrow not feasible

Personal Archive

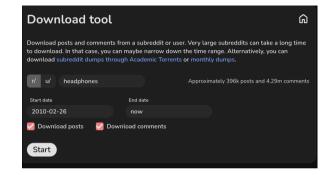
- Continuous collection over weeks
- Impractical: hardware/time, can't capture older posts



(Or building one's own archive over a long period of time like the OP mentioned in another comment, that works too – but it does take time. Though they could load it with data from these archives too if they were so inclined.)

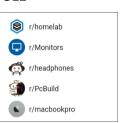
Academic Torrents (Arctic Shift)

- Downloadable historical Reddit datasets
- Good for history + scale



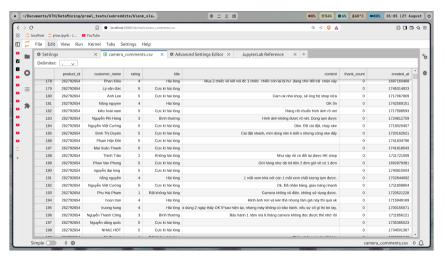
Subreddits Chosen

- A diverse mix of tech and lifestyle communities:
- r/macbookpro, r/GamingLaptops
- r/iphone, r/AppleWatch, r/Monitors, r/ headphones, r/homelab, r/photography
- ...and several others covering home, audio, and PC building.
- total: 134121 posts, 1300190 comments, 2025-06-01 -> 2025-07-31.





Tiki



EDA Highlights

- [Placeholder image: time series of comments/week by subreddit]
- [Placeholder image: length distribution (boxplot or histogram)]
- [Placeholder image: top product mentions (bar) via keyword/NER]
- Takeaways (placeholder): spikes follow launch X; r/homelab comments 2× longer.

Preprocessing Pipeline

Convert 32 JSONL \rightarrow 2 Parquet (Nushell + Polars); keeping as many meaningful columns as possible.

Before

Posts: 106 columns

Comments: 69 columns

After

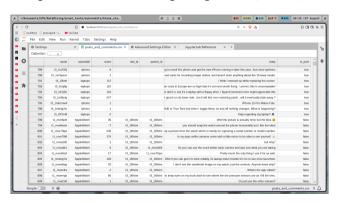
Posts: 28 columns

Comments: 17 columns

• Cleaning:

Removed URLs.

- ► Filtered for English-only content for initial analysis. (only 5 thousands comments(out of 1,3 million), but consider keeping since the model is multilingual)
- Top 30 + 20 randomized posts per sub (seeded) to reduce virality bias. (placeholder)



Context Experiment (Per-Comment SA)

- Parent-context vs per-comment classification compared on small labeled set.
- Result: per-comment avoids 512-token truncation and context pollution.

contextualized_hierarchical_output.json : shows what a comment with the context of it's comment tree looks like.



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Sentiment Analysis (SA)

- Approach: Pretrained transformers over simpler baselines (VADER, TextBlob).
- Model: lxyuan/distilbert-base-multilingual-cased-sentiments-student
 - Why: Good multilingual support, lightweight, strong zero-shot performance.
- Execution: Rented GPU on vast.ai for large-scale inference.
- Next Step: Fine-tune on a domain-specific labeled dataset.

```
"name": "t1_n4cku6n",
   "link_id": "t3_1m5i5lg",
   "parent_id": "t3_1m5i5lg",
   "score": 2,
   "body": "Thats literally me!",
   "contextualized_text": "Thats literally me!",
   "body_positive": 0.5420639514923096,
   "body_neutral": 0.0,
   "body_neutral": 0.0,
   "contextualized_text_positive": 0.5420639514923096,
   "contextualized_text_neutral": 0.0,
   "contextualized_text_neutral": 0.0,
},
```



Topic Modeling (Planned)

- Goal: Discover key themes and issues per subreddit.
- **Method:** Evaluate BERTopic vs. traditional methods (LDA/NMF).
- **Process:** Preprocess with domain stopwords, lemmatization.
- Output: Top topics, representative comments, and trend lines.
- Target: coherence c_v threshold and stable topics across subsamples. (placeholder)
- [Placeholder image: intertopic distance map (BERTopic mock)]
- [Placeholder image: top-words table for 2 sample topics]

CLI Tool (Planned)

- A pipeline for repeatable analysis:
- ingest: Raw data to Parquet.
- clean: Filter and normalize data.
- sentiment: Run batch sentiment analysis.
- topics: Train and apply topic models.
- report: Aggregate results and export.

Risks & Mitigations

- Bias: Sampled randomly across multiple, diverse subreddits.
- **Time Drift:** Included time slices to compare cohorts.
- Reproducibility: Pinned environments, config-driven runs, and stored seeds.
- Ethics: Used public data, followed platform ToS, aggregated results.

Next Steps

- Finalize EDA visualizations.
- Run sentiment analysis at scale.
- Pilot and select a topic modeling approach.
- Build and demo the core CLI workflow.
- Begin labeling for fine-tuning sentiment model.

