



# NLP Sentiment Analysis model

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# What's new ...

- Dataset
  - Amazon Electronics Reviews
    - Label imbalance
    - Lack of diversity
  - X/Reddit API
    - Time or money consuming
  - Goemotions raw data
    - 200,000+ Reddit comments with 27 emotions + neutral
    - P/N → multi-label
- Generating function
  - Train a model
    - Insurmountable Challenges on technique and cost
  - Google Gemini API
    - Fast, controllable, and easy to integrate

	text	labels
0	That game hurt.	['sadness']
1	You do right, if you don't care then fuck 'em!	['neutral']
2	Man I love reddit.	['love']
3	[NAME] was nowhere near them, he was by the Fa...	['neutral']
4	Right? Considering it's such an important docu...	['gratitude']



## Model

	Micro F1	Macro F1	Weighted F1
TF-IDF+LR	0.2505	0.1657	0.2211
BERT + RF	0.3775	0.3164	0.3814

- TF-IDF + Logistic Regression (deployed model)
  - Less accurate
  - Light(less than 700MB) and easy to deploy
- BERT + Random Forest
  - More accurate
  - Large volumn(Around 10GB) → less portable, hard and expensive to deploy



# Implement and Deployment

- Backend
  - Implemented using `TfidfVectorizer` and `LogisticRegression` from `scikit-learn`.
  - Encodes text via `SentenceTransformer`, trained with `RandomForestClassifier` from `scikit-learn`.
  - Built with `Flask` + `flask_cors`, serves `/predict` and `/revise` endpoints via JSON
  - Deployed to Google Cloud Run using `Docker`
- Frontend
  - UI built with `React`, styled with `Ant Design`; uses `axios` for API calls
  - Built with `npm build`, then deployed using Firebase.

# Demo and Takeaways

Demo URL:

<https://nlp-model-83d1c.web.app/>

Future Work:

- Accuracy ↑
  - Filter noisy or short samples to improve label quality
  - Try smaller transformer models
- Size ↓
  - Replace Random Forest with lightweight classifier like MLP
  - Compress embeddings via PCA or use ONNX for quantization

