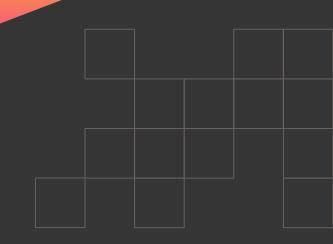
AT&T DESP Internship June 4th, 2025

## Fake News Project



## Project roadmap

model



## **Project Overview**

Problem: The spread of misinformation.

- Fake News breaks down public trust, democracy, and social cohesion.
- Important to develop scalable and intelligent tools to distinguish credible sources from deceptive articles.



## **Project Overview**

# Solution: The Fake News Detection App.

- ML Model trained on hundreds of thousands of datasets, extracting features like title and text.
- Uses NLP and classification pipelines to accurately predict the credibility of an article.
- Simple UI for users to input URLs of articles to determine its legitimacy.



#### Meet the team



Jacob Truong
CI/CD, Frontend,
Machine Learning, and
API Development

Configured CI/CD to allow automatic deployment on AKS, created home page and analysis page elements, developed binary classification model, implemented endpoints to call web scraper and ml model



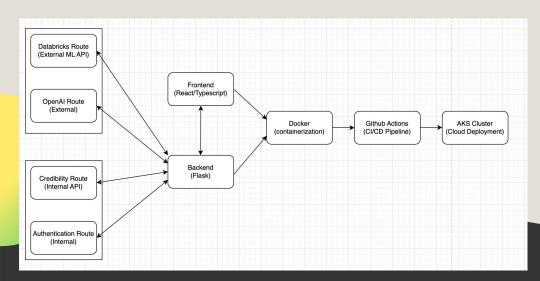
Aaron Luu

Deployment, Data Pipeline,
API Security, Data

Visualization, GenAI

Configured Docker Images to deploy on AKS, Inserted and visualized data on Databricks, implemented JWT, rate limiting, logs of errors, and GenAI use to summarize and assist in analysis of articles.

## Microservices And Architecture



#### Technologies:

- **Frontend**: TypeScript + React
- **Backend**: Flask (Python)
- Authentication: JWT Auth (Flask-JWT-Extended)
- ML Model: Databricks (Spark ML / PySpark)
   Article Analysis: OpenAI GPT via Flask route
- Containerization: Docker
   CI/CD: GitHub Actions
- **Deployment**: Azure Kubernetes Service (AKS)

#### **Endpoints:**

- Docker images built from frontend and backend
- Github actions trigger CI/CD workflow
- Images deployed to AKS
- AKS serves frontend/backend
- Frontend calls backend APIs

### ML Model and Results

#### What was Used?

#### TFIDF:

- Lightweight
- Extracts key term frequencies

#### FastText:

 Captures semantic meaning and word relationships

#### Logistic Regression:

 Provides linear decision boundaries

#### Random Forest:

 Handles non linear patterns and reduces overfitting

#### Deployment and API Interaction

Model is deployed on a serving endpoint with rate limiting on databricks. Returns a prediction and confidence score given text.

#### Limitations

- No GPU
- Memory limitations





Test Set Metrics:

F1 Score (weighted): 0.8008 Precision (weighted): 0.8039

Accuracy: 0.8013

Recall (weighted): 0.8013

AUC (OvR): 0.9068

#### Classification Report:

		precision	recall	f1-score	support	
	0	0.78	0.85	0.81	9943	
	1	0.83	0.75	0.79	9864	
accuracy				0.80	19807	
macro	avg	0.80	0.80	0.80	19807	
weighted	avg	0.80	0.80	0.80	19807	

# Security and Monitoring

#### Protecting endpoints:

#### Rate limiting / Retries

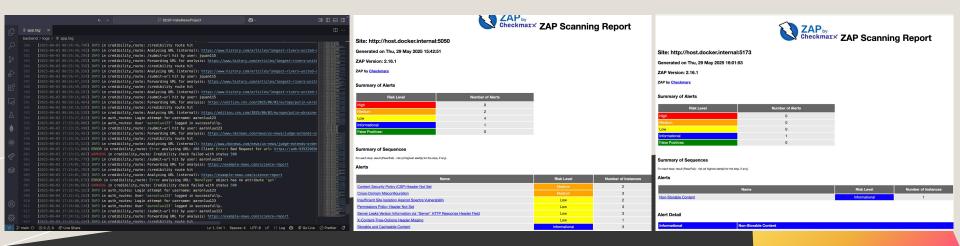
Introduced Flask Limiter to allow up to 100 API calls per min, Also included Tenacity and retries in case of unforeseen errors (network crashes, rate limiting, timeout)

#### **Error Logging**

Included helpful error messages to debug code and troubleshoot problems during API calls. Includes a file App.log to keep track of all logs.

#### **JWTs**

Users create an account with an associated token, every API call is verified with that token.



## GenAI Usage

#### What was Used?

#### ChatGPT-40 / Github Copilot:

- Article summarization
- Credibility analysis assistance
- Code Generation:
  - Data cleanup
- Machine Learning
   Research/Guidance

#### Canva Al

- UI design and starting point

#### Usefulness

- Accelerated UI development with reusable templates
- Helped debug backend and deployment issues

#### Limitations

- Struggled with large codebases or context-heavy debugging
- Sometimes produced overly generic or incorrect suggestions
- Required manual validation of outputs (Postman)







## Retrospective



#### Challenge 1

Limited funding restricted our ability to train large models like BERT or Roberta, which required more compute resources than we could afford.

#### Challenge 2

CPU-only training was slow and inefficient. We lacked access to GPUs and had issues with insufficient and messy data.

### Lessons Learned



#### Lesson 1

Always test code thoroughly before pushing to avoid unnecessary AKS costs.

#### Lesson 2

Ask questions early—waiting until the last minute adds stress and risk.

#### Lesson 3

Use logging and error tracking consistently to debug issues faster, especially in production (Postman to test API calls!)

Thank you everyone, and Thank you to AT&T for making this program possible!

Any questions? Ask away!

