

# **NLP Applications - Assignment 2 – PS-7 (Part A – TASK B)**

## **Group 39**

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## **Problem Definition: PART A - Task B : Enhancement Plan (2 Marks)**

Provide a detailed documentation that would explain the step – by- step process to enhance your Sentiment Analysis Application to develop and expose RESTful APIs that allow external applications to access the sentiment analysis engine.

# Enhancement Plan

## Summary:

**REST APIs for Sentiment Analysis Application** (preprocessing + model inference) through REST endpoints such as:

- POST /api/v1/sentiment → analyze a text string
- POST /api/v1/sentiment/batch → analyze multiple texts
- POST /api/v1/sentiment/file → analyze an uploaded .txt file
- GET /api/v1/health → health check for monitoring

**Details of all enhancements in steps: (Refactor code into a clean service layer)**

1. Restructure code
2. API contracts
3. REST implementation
4. Validation & Security
5. Versioning & Documentation
6. Performance Improvements
7. Logging, Metrics and Error Handling
8. Testing
9. Deployment Plan

## Step 1 : Restructure - UI (HTML) and APIs call the same core inference function.

```
project/app.py          # Flask entry
sentiment/preprocess.py  # cleaning/tokenization/lemmatization
model.py               # load model + predict
service.py             # sentiment_pipeline(text) orchestration
api/routes.py          # API endpoints (Blueprint)
schemas.py             # request/response validation helpers
templates/             # existing UI
static/                # existing UI assets
```

In sentiment/service.py, create a single callable:

- analyze\_text(text: str) -> dict
  - returns: label, confidence, optional scores, cleaned text, etc.

## Step 2 — Define API contract (inputs/outputs)

### 2.1 Request format (JSON)

For single text:

```
{ "text": "I love this product!" }
```

Optional fields you can support:

return\_scores: true/false

language: "en" (if you later add multilingual support)

### 2.2 Response format (JSON)

```
{
  "request_id": "uuid",
  "label": "positive",
  "confidence": 0.91,
  "scores": { "positive": 0.91, "neutral": 0.07, "negative": 0.02 },
  "processing_ms": 12
}
```

### 2.3 Standard error format

```
{
  "request_id": "uuid",
  "error": {
    "code": "INVALID_INPUT",
```

```
        "message": "Field 'text' must be a non-empty string."
    }
}
```

## Step 3 — Implement REST endpoints in Flask using Blueprints

Create a blueprint `api/routes.py` and register it in `app.py` under `/api/v1`.

### Health

GET `/api/v1/health`  
Returns: status OK + model loaded + version  
`{ "status": "ok", "model_loaded": true, "version": "1.0.0" }`

### Single text inference

POST `/api/v1/sentiment`  
Body: `{ "text": "..." }`  
Returns: label + confidence (+ scores optional)

### Batch inference

POST `/api/v1/sentiment/batch`  
Body:  
`{ "texts": ["Good", "Bad", "Okay"] }`  
Returns list of results with indices.

### File upload

POST `/api/v1/sentiment/file`  
multipart/form-data with `file=@input.txt`  
Server reads file content → analyze.

## Step 4 — Add input validation + security defaults

### Validation rules

- text must be string, trimmed, length > 0
- Set max length (example: 5,000 chars) to prevent abuse
- For batch: max items (example: 100)

### Security & hardening

- Disable debug in production
- Add CORS if external frontend/mobile apps will call it
- Add basic API key auth (simple and assignment-friendly)

### **API key approach**

- Client sends: Authorization: Bearer <API\_KEY>
- Server checks it against env var API\_KEY

## **Step 5 — Add versioning and documentation (OpenAPI)**

### **Versioning**

Use /api/v1/... so you can add /api/v2/... later without breaking clients.

### **Documentation**

a simple README.md section with curl examples

## **Step 6 — Performance improvements for API mode**

### **Few points:**

- Load model once at startup, not per request
- Keep preprocessing objects (tokenizer/vectorizer) in memory

## **Step 7 — Logging, monitoring, and error handling**

### **Logging**

- Log per request:
  - request\_id
  - endpoint
  - processing time
  - errors (stack trace only in server logs)

### **Monitoring essentials**

- /health endpoint for uptime checks
- response time tracking (processing\_ms)

## Consistent error handling

Use Flask error handlers for:

- 400 (bad request)
- 401 (unauthorized)
- 500 (server error)

## Step 8 — Testing (very important for “detailed documentation”)

- Unit tests
- preprocessing functions
- model prediction function
- API tests
- Use Flask test client or pytest:
- valid request returns 200 + expected schema
- empty text returns 400 with INVALID\_INPUT
- missing/invalid API key returns 401

### Manual tests (curl examples)

```
curl -X POST http://localhost:5000/api/v1/sentiment \
-H "Content-Type: application/json" \
-H "Authorization: Bearer YOUR_KEY" \
-d '{"text": "This is amazing"}'
```

## Step 9 — Deployment plan

### Local run

- python app.py (or flask run)

### Production run

- Use a WSGI server:
- gunicorn app:app

### Add config via environment variables:

- API\_KEY
- MODEL\_PATH
- MAX\_TEXT\_LENGTH
- LOG\_LEVEL



### **Deploying on Docker:**

- Dockerfile builds image
- container exposes port 5000
- env vars injected at runtime

# Appendix:

## 1. API Endpoint Table

Endpoint	Method	Purpose
/api/v1/health	GET	Health + readiness
/api/v1/sentiment	POST	Single text inference
/api/v1/sentiment/batch	POST	Batch inference
/api/v1/sentiment/file	POST	File upload inference

## 2. Evidence of executing in BITS OSHA Cloud lab

The screenshot shows a web browser window with multiple tabs. The active tab is titled "Sentiment Analyzer" and displays the application's interface. The URL bar shows "http://127.0.0.1:7614". The application has a header "Sentiment Analyzer" and a subtitle "Enter text or upload a .txt file. The app preprocesses and predicts sentiment." Below this, there are two input sections: "Text input" with a text area containing "positive neutral negative" and "OR" below it, and "Upload .txt file" with a "Browse..." button and "No file selected." text. At the bottom of the input section are "Analyze Sentiment" and "Clear" buttons. The "Result" section shows "Neutral" as the predicted sentiment, with "Confidence (proxy): 0.44" and "Compound: -0.0258". Below the result is a section for "Original text" with a placeholder for a text file icon.

This screenshot shows the same Sentiment Analyzer application, but with a more detailed result breakdown. The "Result" section shows "Neutral" as the predicted sentiment, with "Confidence (proxy): 0.44" and "Compound: -0.0258". Below this, there are two sections: "Original text" and "Preprocessed text", both showing the input text "positive neutral negative". The "Sentiment score breakdown" section features a bar chart showing the scores for "positive", "neutral", and "negative" sentiments. The "positive" bar is the highest, followed by "negative", and then "neutral". The chart is titled "Sentiment score breakdown" and has a legend indicating "Score".

Sentiment	Score
positive	0.44
neutral	-0.0258
negative	0.44

Course / Experiment:

Date:

Slot:

Book

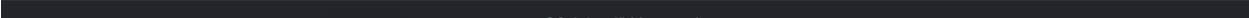
--- Select Course / Experiment ---

--- Select Date ---

--- Select Slot ---

A fresh AWS instance was launched for you. Please wait for a few minutes before connecting.

Course (Virtual Lab)	Machine	Time Used	Time Remaining	State	Control
25S1NSP2-13-NLPA(Virtual Lab)	i-07f41f29f503de47f	16m 38s	21h 43m	started	<div>Connect</div> <div>Stop</div>



### 3. Design & Flows

