Subject: Deployment of ML models (Assignment 2) Team Name: Mlops Squad

Team members

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Assignment 2:

Email Spam Detection. Classify emails as spam or not spam using NLP techniques.

Dataset: SMS Spam Collection (Text Classification)

https://www.kaggle.com/datasets/thedevastator/sms-spam-collection-a-more-diversedataset

Task 1:

- Create a ML model.
- Build REST API for best_model_parameter[method to be used GET], prediction[method to be used POST] and training[method to be used POST]

Task 2:

- Use hyper-parameter tuning technique to find the best model and log all experiment run using ML-Flow.
- All the Experiment artifact should be visible in the code github repository

Task 3:

- Create a Docker container for the Backed API and push it to docker hub
- A working Docker Image of your application showcasing both the ML-flow UI and the REST API

Solution:

Spam Detection API

This project implements a REST API for classifying emails as spam or not spam using NLP techniques.

Features

- Train multiple models (Decision Tree, Logistic Regression, Random Forest, Multinomail Naive Bayes).
- Hyperparameter tuning with MLflow logging.
- REST API endpoints for model parameters, training, and prediction.

- Dockerized application with MLflow UI.

Setup

1. Project Structure (spam-message-detection):

```
MLOPS SQUAD/
                         # Flask REST API application
 ⊢— app.py
 — spam classifier.py
                         # Script for training models and hyperparameter tuning
 — Dockerfile
                         # Dockerfile for containerizing the application
                         # List of Python dependencies
 — requirements.txt
 README.md
                         # Project documentation
 — dataset/
                         # Folder for dataset
  └─ spam.csv
                         # Dataset file |
 — models/
                        # Folder for trained models
  ☐ spam model.pkl # Best model saved using pickle
   └─ vectorizer.pkl
                        # Vectorized model
                        # MLflow experiment logs (generated after running mlflow)
  — mlruns/
```

Clone the repository:

- git clone git@github.com:Rajesh-Lohani/mlops squad.git
- cd mlops_squad

File Descriptions

1. app.py

Purpose: Implements the Flask REST API.

Endpoints:

/best_model_parameters (GET): Returns the best model parameters.

/train (POST): Trains the model with new data.

/predict (POST): Predicts whether a message/email is spam or not.

Dependencies: Loads the best model (spam_model.pkl) and vectorizer.pkl.

2. spam_classifier.py

Purpose: Script for training multiple models, performing hyperparameter tuning, and logging experiments with MLflow.

Outputs:

Saves the best model as spam_model.pkl.

Logs experiments in the mlruns/ folder.

3. Dockerfile

Purpose: Defines the Docker image for the application.

4. requirements.txt

Purpose: Lists all Python dependencies.

5. README.md

Purpose: Provides an overview of the project, setup instructions, and usage details.

How to run the application using Docker image – available in docker hub: 1. Pull the docker image using below command: docker pull rajeshlohani/mlops_squad:latest 2. Run the Docker Container: docker run -p 5000:5000 -p 5001:5001 rajeshlohani/mlops squad 3. Applications: Mlflow UI url : http://localhost:5001/ REST API url: http://localhost:5000/ 4. Endpoints a. GET /best_model_parameters: Get the best model's hyperparameters. Sample url: http://127.0.0.1:5000/best_model_parameters Output: "best_model_parameters": { "alpha": 0.1, "class_prior": null, "fit_prior": true, "force alpha": "warn" } b. POST /train: Train the model with new data. Sample url : http://127.0.0.1:5000/train Input dataset: "sms": ["Free entry in 2 a wkly comp", "Hey there, how are you?", "Win a brand new car!", "Can we meet tomorrow?"], "label": ["spam", "ham", "spam", "ham"] Output: "accuracy": 96.0,

"message": "Model trained successfully"

c. POST /predict: Predict whether a given SMS is spam or not.

```
Sample url: <a href="http://127.0.0.1:5000/predict">http://127.0.0.1:5000/predict</a>
Input body:

{
    "sms":"Hello! you won 1000"
}

Output:

{
    "prediction": "spam"
```