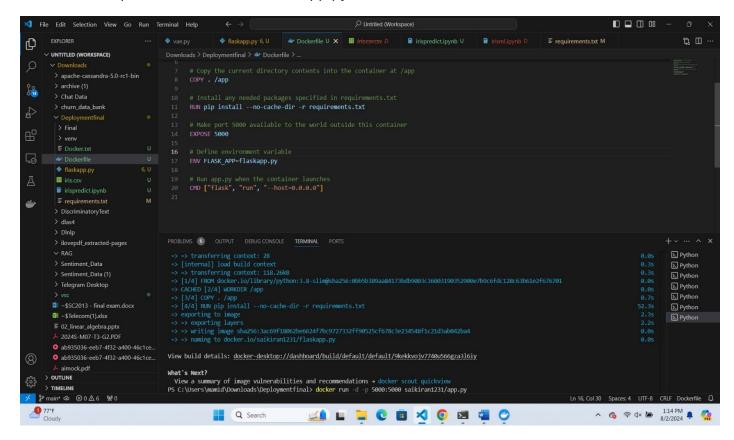
Github url: https://github.com/MSK-hash/Final.git

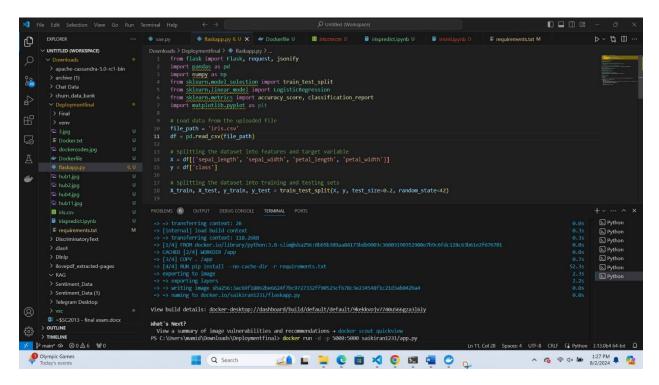
Saikiran\_Mamidala-500209412

Provide the Docker commands used along with screenshots:

docker build -t saikiran1231/app.py.

docker run -d -p 5000:5000 saikiran1231/app.py





This Python script is a Flask web application that loads a machine learning model to classify iris flowers based on their features. The application provides two main functionalities:

- 1. Displaying the model's accuracy and classification report.
- 2. Predicting the class of an iris flower based on user-provided features.

## **Detailed Explanation**

## 1. Import Libraries:

 The script imports necessary libraries for data manipulation (pandas), numerical operations (numpy), machine learning (sklearn), visualization (matplotlib), and web development (flask).

#### 2. Load Dataset:

o The Iris dataset is loaded from a file named iris.csv using pandas.

### 3. Data Preparation:

- o The dataset is split into features (X) and the target variable (y).
- The features include sepal\_length, sepal\_width, petal\_length, and petal\_width.
- The target variable is the class of the iris flower.

#### 4. Split Data:

 The dataset is divided into training and testing sets using train\_test\_split from sklearn. This ensures that the model can be trained and evaluated on separate data.

#### 5. Train Model:

 A Logistic Regression model is instantiated and trained on the training data (X\_train, y\_train).

#### 6. Make Predictions:

o The trained model is used to make predictions on the test data (X\_test).

#### 7. Evaluate Model:

- o The accuracy of the model is calculated using accuracy\_score.
- o A detailed classification report is generated using classification\_report.

## 8. Flask Web Application:

o The Flask app is initialized and configured to run on host 0.0.0.0 and port 80.

## 9. **Home Route (/)**:

 The home route (/) returns the model's accuracy and classification report in HTML format.

### 10. Prediction Route (/predict):

- The prediction route (/predict) accepts POST requests with JSON data containing the features of an iris flower.
- It extracts the features from the JSON data, reshapes them into the required format, and uses the trained model to predict the class of the iris flower.
- The predicted class is returned as a JSON response.

Screenshots of your application running in the container:

#### 1. Dockerfile:

 The Dockerfile defines the instructions to build the Docker image, including the base image (Python 3.9-slim), setting the working directory, copying the application files, installing dependencies, and exposing the necessary port.

# 2. Building the Docker Image:

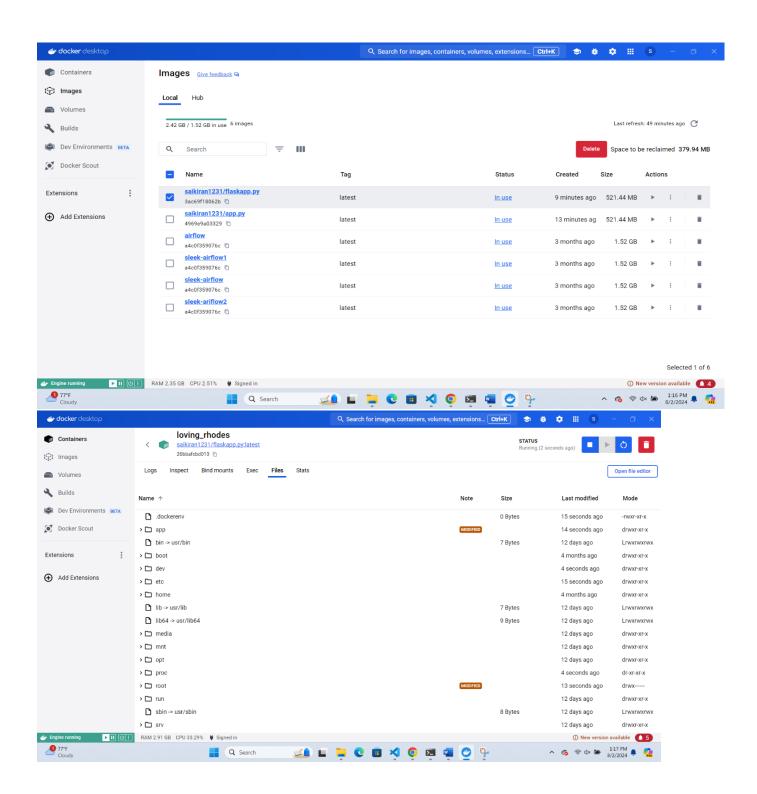
The Docker image is built using the command docker build -t iris-classifier.,
which packages the application and its dependencies into an image.

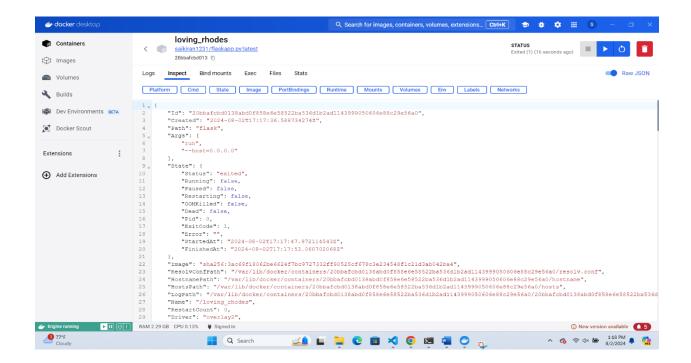
## 3. Running the Docker Container:

 The application is run inside a Docker container using the command docker run -p 4000:80 iris-classifier. This maps port 80 inside the container to port 4000 on the host machine, allowing access to the application via http://localhost:4000.

## 4. Accessing the Application:

Once the container is running, users can access the application in their web browser. The home page displays the model's accuracy and classification report. Additionally, the /predict endpoint can be used to make predictions by sending POST requests with iris flower features.





#### Docker in Hub:

# **Pull the Docker Image from Docker Hub**

First, ensure that you have Docker installed on your machine. Then, pull your Docker image from Docker Hub

#### **Run the Docker Container**

Run a container from the pulled Docker image

By containerizing the application with Docker, we ensure a reliable and reproducible deployment process, making it easier to manage and scale the application in different environments, including local development, staging, and production.

