



MLOPS-Assignment-Group-41- / readme.md



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7b2a55b · 26 minutes ago



307 lines (190 loc) · 8.01 KB

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Heart Disease Prediction System – MLOps

Course: MLOps

Assignment: End-to-End MLOps Pipeline

Dataset: UCI Heart Disease Dataset

Group: Group 41

Repository: <https://github.com/rahulvg/MLOPS-Assignment-Group-41->

Problem Statement

The objective of this project is to design, develop, and deploy a scalable, reproducible, and production-ready machine learning system to predict the presence of heart disease based on patient health attributes.

The solution follows modern MLOps best practices, including experiment tracking, CI/CD automation, containerization, Kubernetes deployment, and monitoring.

1. Setup and Installation Instructions

1.1 Local Environment

Python Version: 3.10

Install Dependencies

```
pip install -r requirements.txt
```



Run Training File(Runs experiment with different parameter on Logistics Regression and Random forest)

```
python train/train_experiment.py
```



Run all unit tests using Pytest

```
pytest
```



Create report using pytest

```
pytest --html=pytest_report.html
```



Launch MLflow UI (Local SQLite DB)

```
mlflow ui --backend-store-uri sqlite:///mlflow.db
```



Access MLflow at:

```
http://localhost:5000
```



1.2 Verification of Docker Build and Execution via GitHub Actions

Due to organizational restrictions that prevent local installation of Docker Desktop, the Docker image build and container execution were verified using **GitHub Actions**, which provides a Docker-enabled runner environment.

This ensures that containerization and execution are **reproducible, verifiable, and independent of local system constraints**.

1. Navigate to the GitHub repository:

<https://github.com/rahulvg/MLOPS-Assignment-Group-41->

2. Click on the **Actions** tab in the repository.

3. Select the most recent workflow run under the **CI pipeline**.

4. Open the workflow run and inspect the following steps:

- **Build Docker image**

This step executes the Docker build command using the project's `Dockerfile`.

- **Run Docker container and test API**

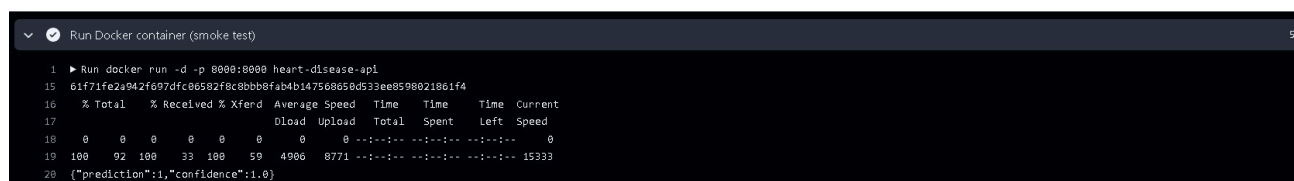
This step starts the container and invokes the `/predict` endpoint using a sample JSON request.

Evidence of Successful Docker Execution

Within the GitHub Actions workflow logs, the following evidence can be observed:

- Docker build logs confirming successful image creation
- Container startup logs indicating the FastAPI service is running
- Successful HTTP response from the `/predict` endpoint returning a prediction and confidence score

Screenshot of successful Docker run event



The screenshot shows a terminal window titled "Run Docker container (smoke test)". The command executed is `Run docker run -d -p 8000:8000 heart-disease-api`. The output shows a successful container startup and a successful HTTP response from the `/predict` endpoint. The response is a JSON object: `{"prediction":1,"confidence":1.0}`.

```
1 ▶ Run docker run -d -p 8000:8000 heart-disease-api
15 61f71fe2a042f97dfc86582f8c8bbbfab4b147568650d53ee8598021861f4
16 % Total    % Received % Xferd  Average Speed   Time    Time     Time  Current
17          %             Dload  Upload  Total   Spent    Left    Speed
18  0     0     0    0     0     0      0      0 --:--:-- --:--:-- --:--:--    0
19 100   92 100   33 100   59   4906   8771 --:--:-- --:--:-- --:--:-- 15333
20 {"prediction":1,"confidence":1.0}
```

1.3 Kubernetes (Local Deployment with Minikube)

Start Minikube

```
minikube start --container-runtime=containerd
```



```
Administrator: Windows PowerShell
PS E:\RGI3\MLOPS> kubectl get pods
NAME                                READY   STATUS    RESTARTS   AGE
heart-disease-api-9c4f666d8-wgxrw   1/1     Running   0           15s
PS E:\RGI3\MLOPS> kubectl get svc
NAME                TYPE        CLUSTER-IP    EXTERNAL-IP   PORT(S)          AGE
heart-disease-service  NodePort    10.110.136.89 <none>        80:30007/TCP     15s
kubernetes           ClusterIP   10.96.0.1     <none>        443/TCP          8m21s
PS E:\RGI3\MLOPS> minikube service heart-disease-service
```

NAMESPACE	NAME	TARGET PORT	URL
default	heart-disease-service	80	http://172.21.82.246:30007

```
* Opening service default/heart-disease-service in default browser...
PS E:\RGI3\MLOPS>
```

Build Docker Image Inside Minikube

```
minikube image build -t heart-disease-api .
```



Deploy Application

```
kubectl apply -f k8s/deployment.yaml
kubectl apply -f k8s/service.yaml
```



Expose Service

```
minikube service heart-disease-service
```



Two models were evaluated:

- **Logistic Regression** – chosen as a strong, interpretable baseline for structured medical data
- **Random Forest** – included to capture non-linear relationships and feature interactions

All numerical features were standardized using **StandardScaler**, and preprocessing was implemented through a unified **scikit-learn Pipeline** to ensure reproducibility, prevent data leakage, and enable deployment-safe inference.

Hyperparameter Tuning

- Logistic Regression: $C \in \{0.1, 1.0, 10.0\}$
- Random Forest:
 - `n_estimators` $\in \{100, 200\}$
 - `max_depth` $\in \{\text{None}, 10\}$

Each configuration was logged as a separate experiment using **MLflow**.

Evaluation

Models were evaluated using **5-fold cross-validation** with the following metrics:

- Accuracy
- Precision
- Recall
- ROC-AUC

Final Model

Logistic Regression with $C = 0.1$ was selected due to:

- Consistent cross-validation performance
- Lower variance across folds
- Better generalization
- Simpler and more interpretable behavior

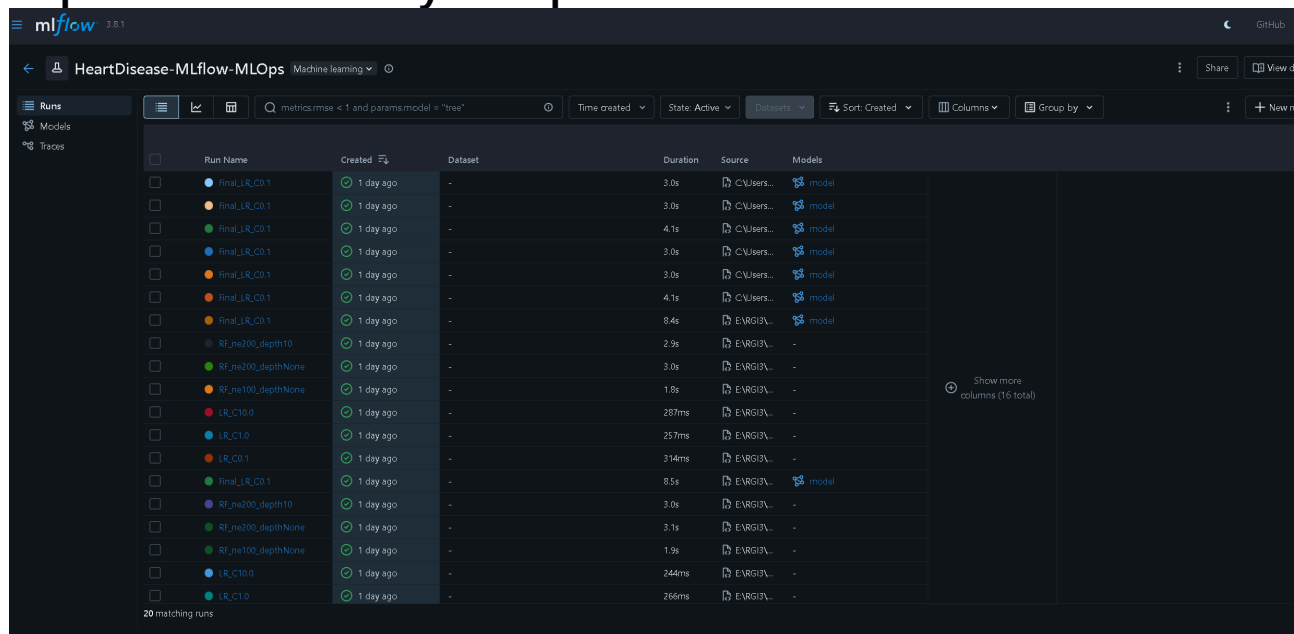
Its stability and ease of monitoring make it well-suited for a production-oriented MLOps pipeline.

3. Experiment Tracking

MLflow was integrated to track:

- Model parameters
- Cross-validation metrics
- Model artifacts

All experiments are logged under a dedicated MLflow experiment for easy comparison.

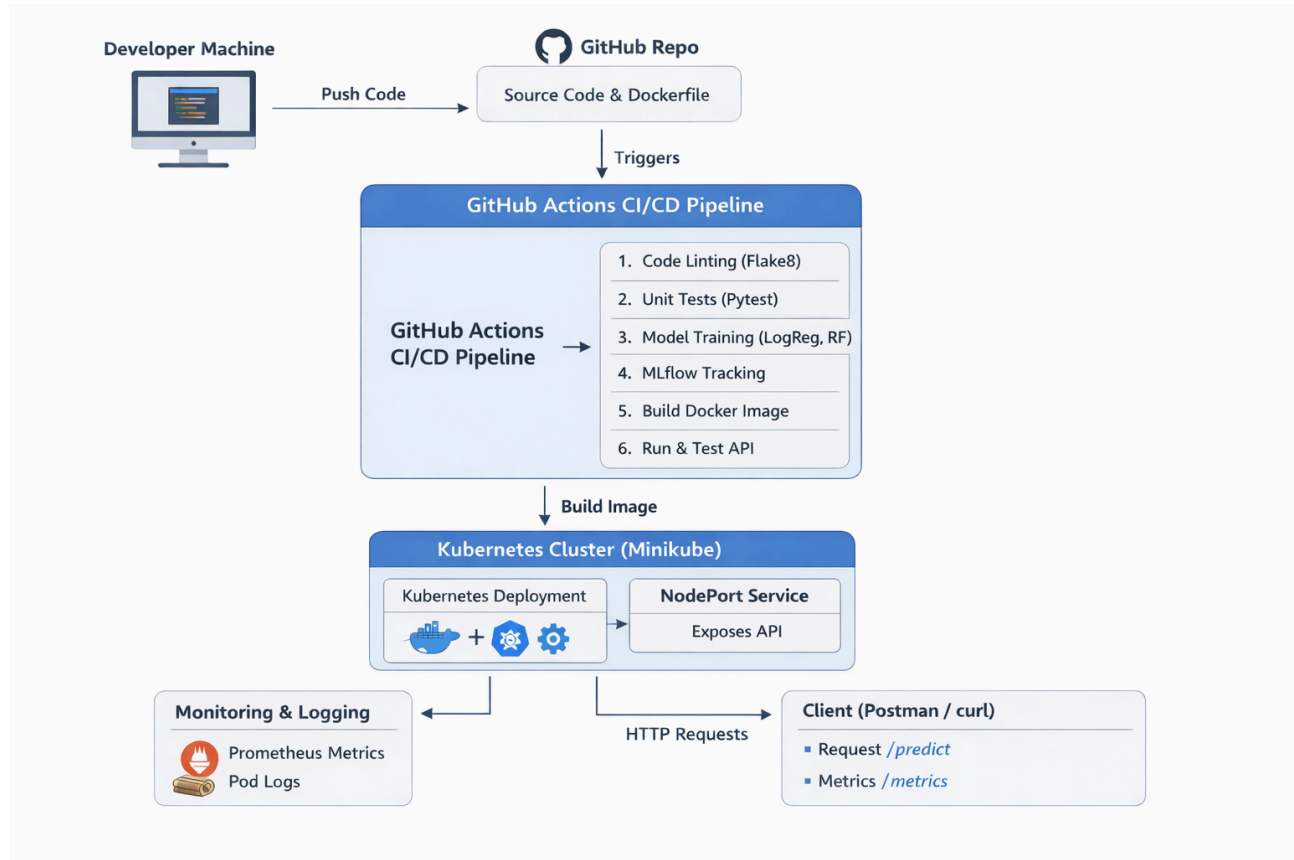


Run Name	Created	Dataset	Duration	Source	Models
final_LR_C0.1	1 day ago	-	3.0s	C:\Users...	model
final_LR_C0.1	1 day ago	-	3.0s	C:\Users...	model
final_LR_C0.1	1 day ago	-	4.1s	C:\Users...	model
final_LR_C0.1	1 day ago	-	3.0s	C:\Users...	model
final_LR_C0.1	1 day ago	-	3.0s	C:\Users...	model
final_LR_C0.1	1 day ago	-	4.1s	C:\Users...	model
final_LR_C0.1	1 day ago	-	8.4s	ENRGISL...	model
RF_ne200_depth10	1 day ago	-	2.9s	ENRGISL...	-
RF_ne200_depthNone	1 day ago	-	3.0s	ENRGISL...	-
RF_ne100_depthNone	1 day ago	-	1.8s	ENRGISL...	-
LR_C10.0	1 day ago	-	287ms	ENRGISL...	-
LR_C1.0	1 day ago	-	257ms	ENRGISL...	-
LR_C0.1	1 day ago	-	314ms	ENRGISL...	-
final_LR_C0.1	1 day ago	-	8.5s	ENRGISL...	model
RF_ne200_depth10	1 day ago	-	3.0s	ENRGISL...	-
RF_ne200_depthNone	1 day ago	-	3.1s	ENRGISL...	-
RF_ne100_depthNone	1 day ago	-	1.9s	ENRGISL...	-
LR_C10.0	1 day ago	-	244ms	ENRGISL...	-
LR_C1.0	1 day ago	-	266ms	ENRGISL...	-

Model Packaging and Reproducibility

- Final model saved as a serialized scikit-learn Pipeline
- Model can be found at **final_model\heart_disease_lr_c01.pkl** in git repo.
- Preprocessing included within the model
- Reproducible inference guaranteed
- Dependencies listed in requirements.txt
- Artifacts stored and versioned using MLflow check **mlflow_experiment.db** in git repo

4. Architecture Diagram



5. CI/CD Pipeline

Tools Used

- GitHub Actions
- Pytest
- Flake8
- Docker

Pipeline Stages

- Code linting
- Unit testing
- Model training
- Docker image build
- API smoke testing

Workflow run deleted successfully.

Actions New workflow

All workflows

MLOps CI Pipeline

Management

- Caches
- Attestations
- Runners
- Usage metrics
- Performance metrics

All workflows

Showing runs from all workflows

Filter workflow runs

23 workflow runs

Event	Status	Branch	Actor	Time
Docker	Success	mlops_assignment	rahulvg	Dec 31, 2025, 8:58 PM GMT+5:30
Docker	Success	mlops_assignment	rahulvg	Dec 31, 2025, 6:42 PM GMT+5:30
Docker	Success	mlops_assignment	rahulvg	Dec 31, 2025, 6:38 PM GMT+5:30
Docker	Success	mlops_assignment	rahulvg	Dec 31, 2025, 6:20 PM GMT+5:30
Merge pull request #5 from rahulvg/mlops_assignment	Success	main	rahulvg	Dec 30, 2025, 9:10 PM GMT+5:30
Renamed requirement.txt	Success	init-commit	rahulvg	Dec 30, 2025, 9:08 PM GMT+5:30

Summary

All jobs

- build-test-train

Run details

- Usage
- Workflow file

build-test-train

succeeded yesterday in 1m 44s

Search logs

- Checkout code (1s)
- Set up Python (8s)
- Install dependencies (34s)
- Run linting (flake8) (1s)
- Run unit tests (9s)
- Upload Pytest HTML report (1s)
- Train final model (9s)

```

1 ▶ Run export PYTHONPATH=$(pwd)
12 Registered model 'HeartDiseaseClassifier' already exists, creating a new version of this model...
13 Created version '4' of model 'HeartDiseaseClassifier'.
14
15 ===== Logistic Regression Experiments =====
16
17 Run: LR_C0_1
18 ACCURACY | Mean: 0.8417 | Std: 0.0347
19 PRECISION | Mean: 0.8516 | Std: 0.0648
20 RECALL | Mean: 0.7987 | Std: 0.0277
21 ROC_AUC | Mean: 0.9164 | Std: 0.0213
22
23 Run: LR_C1_0
24 ACCURACY | Mean: 0.8482 | Std: 0.0380
25 PRECISION | Mean: 0.8679 | Std: 0.0586
26 RECALL | Mean: 0.7910 | Std: 0.0174
27 ROC_AUC | Mean: 0.9111 | Std: 0.0196
28
29 Run: LR_C10_0
30 ACCURACY | Mean: 0.8350 | Std: 0.0235
  
```

6. Code Repository

<https://github.com/rahulvg/MLOPS-Assignment-Group-41->

7. Containerization and Deployment

7.1 Dockerized API

- FastAPI-based service
- /predict endpoint
- Accepts JSON input

- Returns prediction and confidence score

7.2 Kubernetes Deployment

- Local Kubernetes using Minikube
- Deployment and NodePort Service manifests
- API tested using curl and Postman

7. Monitoring and Logging

7.1 Logging

- Request-level logging implemented via FastAPI middleware
- Logs include endpoint, HTTP status, and latency
- Logs accessible via Kubernetes pod logs

7.2 Monitoring

- Prometheus-compatible /metrics endpoint exposed
- Metrics include request count and request latency
- Ready for Prometheus and Grafana integration

```

Administrator: Windows PowerShell
heart-disease-api-9cd666d8-fr998 1/1 Running 0 2m19s
PS E:\V8013\ML\OPS> kubectl logs heart-disease-api-9cd666d8-fr998
/usr/local/lib/python3.10/site-packages/sklearn/base.py:348: InconsistentVersionWarning: Trying to unpickle estimator StandardScaler from version 1.0.0 when using version 1.3.2. This might lead to breaking code or invalid results. Use at your own risk. For more info please refer to https://scikit-learn.org/stable/model_persistence.html#security-maintainability-limitations
warnings.warn(
/usr/local/lib/python3.10/site-packages/sklearn/base.py:348: InconsistentVersionWarning: Trying to unpickle estimator LogisticRegression from version 1.0.0 when using version 1.3.2. This might lead to breaking code or invalid results. Use at your own risk. For more info please refer to https://scikit-learn.org/stable/model_persistence.html#security-maintainability-limitations
warnings.warn(
/usr/local/lib/python3.10/site-packages/sklearn/base.py:348: InconsistentVersionWarning: Trying to unpickle estimator Pipeline from version 1.0.0 when using version 1.3.2. This might lead to breaking code or invalid results. Use at your own risk. For more info please refer to https://scikit-learn.org/stable/model_persistence.html#security-maintainability-limitations
warnings.warn(
INFO: Started server process [1]
INFO: Waiting for application startup.
INFO: Application startup complete.
INFO: Uvicorn running on http://0.0.0.0:8080 (Press CTRL-C to quit)
/usr/local/lib/python3.10/site-packages/sklearn/base.py:465: UserWarning: X does not have valid feature names, but StandardScaler was fitted with feature names
warnings.warn(
2025-12-31 14:36:55,174 | INFO | POST /predict | status=200 | latency=0.003s
INFO: http://localhost:8080/POST /predict content-type=application/json http://11.11.11.11 200 OK
PS E:\V8013\ML\OPS>

```

8. Architecture Overview

Client (Postman / curl)

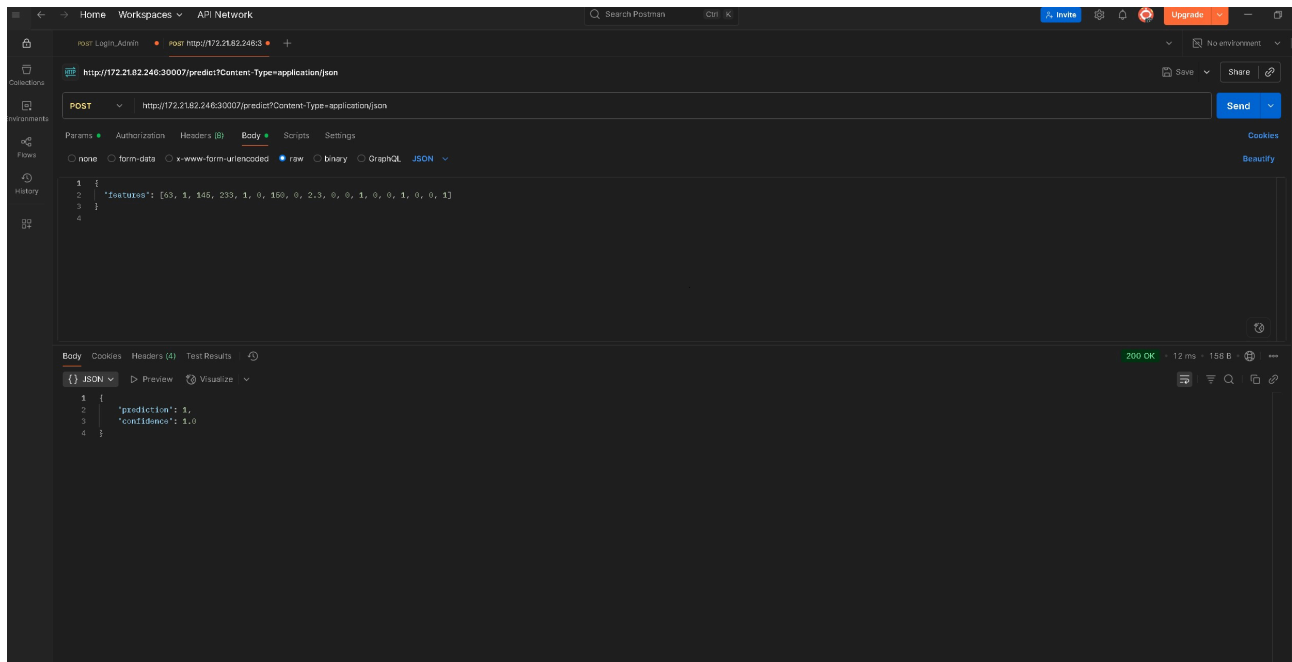
- FastAPI API (/predict, /metrics)
- Scikit-learn Pipeline
- Kubernetes Pod
- NodePort Service

CI/CD is handled using GitHub Actions, and experiment tracking is handled using MLflow.

9 Run using CURL/Postman API

Curl Command

```
"curl -X POST http://127.0.0.1:30007/predict \  
-H "Content-Type: application/json" \  
-d '{"features": [63,1,145,233,1,0,150,0,2.3,0,0,1,0,0,1,0,0,1]}'"
```



Conclusion

This project demonstrates a complete, production-grade MLOps workflow covering data analysis, model development, experiment tracking, CI/CD automation, containerization, Kubernetes deployment, and monitoring.

The system is scalable, reproducible, and aligned with real-world MLOps practices.

Demo video

<https://github.com/rahulvg/MLOPS-Assignment-Group-41/blob/main/Demo%20Video.mp4>

<https://drive.google.com/drive/folders/1hGcu4oyM3TusMy8vnackoOLpug7LhJdP?usp=sharing>

Appendix: Useful Commands

Launch MLflow with Custom Local DB

```
mlflow ui --backend-store-uri sqlite:///E:/RGI3/MLOPS/mlflow.db
```



Rebuild and Redeploy on Minikube

```
minikube image build -t heart-disease-api .  
kubectl delete deployment heart-disease-api  
kubectl apply -f k8s/deployment.yaml
```

