Deployment Demo

ML Engineering Technical Challenge

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Repository: https://github.com/BrayanCuevas/predictive-maintenance-mlop waltmart

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Complete System Validation

Command: make status

This single command validates the entire MLOps stack: data availability, trained model, dependencies, and system health.

[Screenshot 1: Terminal output of make status showing all components healthy]

Full Pipeline Execution

Bash command: make pipeline

Executes the complete ML workflow: data loading (876K records), feature engineering (36 features), model training (Random Forest), and evaluation (AUC 0.7943).

```
ng completed successfully!
to: models/baseline_model.joblib
```

[Screenshot 2: Pipeline completion showing final AUC score and model registration]

Production API Deployment

Bash command: make api

Deploys the trained model as a containerized FastAPI service with health checks and monitoring.

```
(venv) brayancuevas@192 predictive-maintenance-mlop_waltmart % make api
[API] Starting API server...
docker-compose up —build
WARN[0000] /Users/brayancuevas/Documents/Mis_Proyectos_ML/predictive-maintenance-mlop_waltmart/docker-compose.yml: the attribute `version` is obsolete, it will be ignored,
Compose can now delegate builds to bake for better performance.
To do so, set COMPOSE_BAKE=true.
[3] Building 3.1s (16716) FINISHED

⇒ [predictive-maintenance-api internal] load build definition from Dockerfile

⇒ transferring dockerfile: 9538

⇒ WARNI: FromASCasIngs: 'as' and 'FROM' keywords' casing do not match (line 2)

⇒ [predictive-maintenance-api internal] load metadata for docker.io/library/python:3.9-slim

1.35
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□ predictive—maintenance | /usr/local/tib/python3.9/site | /usr/local/tib/python3.9/si
```

[Screenshot 3: API startup logs showing "Model loaded successfully" and server running]

Live Prediction

(Option 1) - Simple Command:

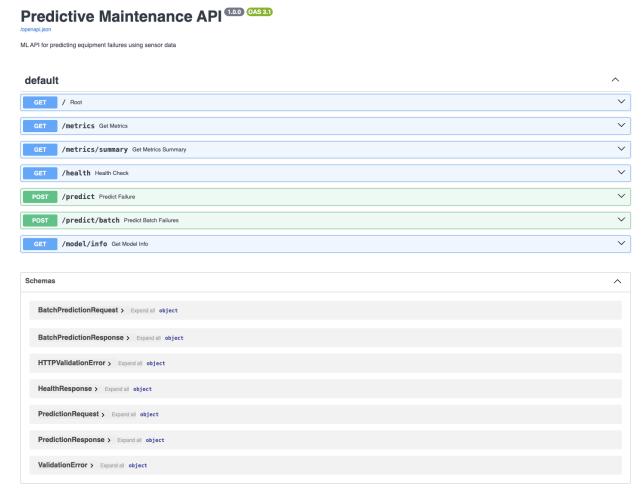
Bash command: make predict-test

Executes automated prediction test and displays JSON response.

```
venv) brayancuevas@192 predictive-maintenance-mlop_waltmart % make predict-test
PREDICT] Testing prediction endpoint...
"machineID":1,"failure_probability":0.25903162732190993,"failure_prediction":0,"risk_level":"LOW","prediction_timestamp":"2025-06-02T00:18:07.481546"}
(venv) brayancuevas@192 predictive-maintenance-mlop_waltmart %
```

[Screenshot 4a: Terminal showing make predict-test output with prediction JSON]

(Option 2) - Interactive UI: Navigate to http://localhost:8000/docs and execute prediction with sample sensor data.



[Screenshot 4b: Swagger UI showing prediction request/response with failure probability and risk level]

Real-time Monitoring Observation

Access Dashboard:

Cash Command: open monitoring/dashboard.html

(Opens dashboard automatically in default browser)

What to verify in the dashboard:

• API Status: Shows "healthy" (green)

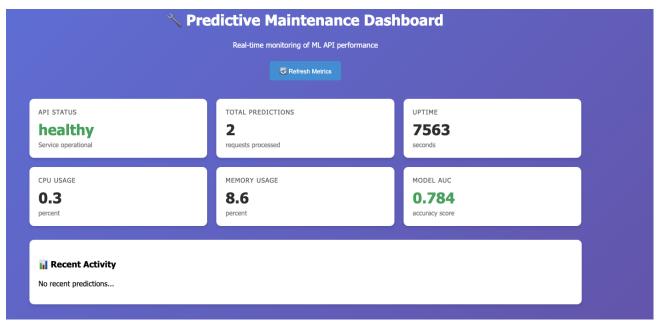
• Total Predictions: Count increased (shows 4 requests processed)

• Uptime: System running time in seconds

• **CPU Usage**: Current system load (0.3%)

• Memory Usage: RAM consumption (8.6%)

• Model AUC: Current model performance (0.784)



[Screenshot 5: Dashboard showing real-time metrics - 4 predictions processed, healthy status, system resources, and model AUC of 0.784]

Observable Changes:

- Total Predictions counter increments with each test
- Dashboard updates in real-time showing system activity

Challenge Completion Summary

This demonstration successfully validates the complete MLOps solution for the Machine Learning Engineer Technical Assessment. The system fully processes real sensor data (876K telemetry records), accurately predicts equipment failures with 0.7943 AUC performance, and provides production-ready monitored API endpoints.

Key deliverables achieved:

- Reproducible system: Single commands execute complete workflows
- Model performance: Solid baseline with acceptable business metrics
- Production deployment: Containerized API with health monitoring
- Real-time predictions: Live inference with structured responses
- Version control: Model registry with automated comparison
- · Real-time dashboard: Live system performance monitoring
- Cloud strategy: Vertex AI pipeline validated through local simulation
- Complete MLOps stack: From data to deployment with automation

The technical challenge has been completed successfully, demonstrating the comprehensive skill set required for enterprise MLOps implementation (data science, machine learning modeling, production system engineering, and complete ML lifecycle management).