

# AB InBev MLOps Challenge

## Deploying a Machine Learning Model as an API

Cristhian Guerrero

MLOps Computer Engineer

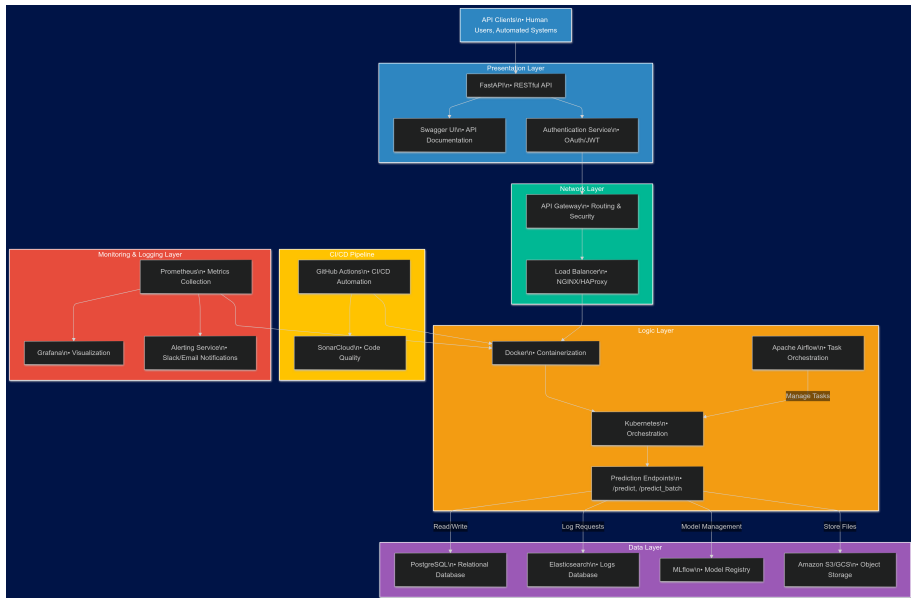
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# Table of Contents

- 1 Introduction
- 2 Architecture
- 3 Development Process
- 4 Challenges and Solutions
- 5 Key Learnings and Findings
- 6 Testing and Quality Assurance
- 7 Results
- 8 Conclusion
- 9 Contact Information

- **Objective:** Deploy a pre-trained machine learning model as a RESTful API.
- **Requirements:**
  - Follow OOP standards in Python.
  - Include inline documentation.
  - Implement batch predictions.
  - Store predictions in a database.
  - Use Docker Compose for local deployment.
  - Run tests and integrate SonarCloud.

# System Architecture



- **FastAPI:** Serves the ML model via RESTful API.
- **PostgreSQL:** Stores predictions made by the model.
- **Docker & Docker Compose:** Containerization and orchestration.
- **GitHub Actions & SonarCloud:** CI/CD pipeline and code quality analysis.

- Analyzed project requirements.
- Designed system architecture.
- Selected appropriate technologies.

- **Python:** Core programming language.
- **FastAPI:** For building the API.
- **SQLAlchemy:** ORM for database interactions.
- **Docker & Docker Compose:** For containerization.
- **GitHub Actions:** For continuous integration.
- **SonarCloud:** For code quality assurance.

# Implementation Steps

- ➊ Set up project structure.
- ➋ Implemented the ML model exposure via FastAPI.
- ➌ Ensured OOP principles by creating classes like `Predictor`.
- ➍ Added inline documentation in English.
- ➎ Enabled batch predictions.
- ➏ Integrated database using SQLAlchemy.
- ➐ Wrote tests and configured SonarCloud.
- ➑ Containerized the application with Docker and Docker Compose.



# Challenge 1: Docker Compose Configuration

- **Issue:** Networking and dependency management between services.
- **Solution:**
  - Defined services in `docker-compose.yml` with proper dependencies.
  - Used environment variables for configuration.

## Challenge 2: SonarCloud Integration

- **Issue:** Configuring SonarCloud with a private GitHub repository.
- **Solution:**
  - Generated and securely stored SonarCloud tokens.
  - Updated GitHub Actions workflow to include SonarCloud analysis.

# Challenge 3: Batch Predictions and Data Integrity

- **Issue:** Handling large volumes of data efficiently.
- **Solution:**
  - Optimized data processing using pandas.
  - Ensured atomic transactions when storing predictions.

# Technical Skills Enhanced

- Deepened knowledge of FastAPI and asynchronous programming.
- Improved understanding of Docker networking and orchestration.
- Gained experience with CI/CD pipelines and code quality tools.

# Best Practices Adopted

- Implemented clean code principles and OOP standards.
- Used environment variables for configuration management.
- Ensured code readability through inline documentation.

- Wrote unit tests for API endpoints using pytest.
- Validated data inputs and outputs.
- Tested database interactions.

- Configured GitHub Actions to run tests on each push.
- Integrated SonarCloud for code analysis.
- Maintained high code coverage and minimal code smells.

# Functional Application

- Met all specified requirements.
- Successfully deployed the API locally using Docker Compose.
- Stored predictions in PostgreSQL database.



## AB InBev MLOps Challenge API 0.1.0 QAS 3.1

[OpenAPI spec](#)

API for predicting customer clusters based on RFM analysis.

### default

GET / Read Root

POST /predict Predict

Endpoint to predict clusters for input data and store predictions in the database.

Args: data (List[InputData]): List of input data instances. db (SessionLocal): Database session.

Returns: dict: Dictionary containing the predicted clusters.

Parameters

No parameters

Try it out

Request body required

application/json

Example Value | Schema

```
{
  "Frequency": 0,
  "MonetaryValue": 0
}
```

Responses

Code	Description	Links
200	Successful Response	No links

Media type

- Successfully deployed a machine learning model as an API.
- Implemented best practices in software development.
- Enhanced skills in MLOps and DevOps tools.

# Future Improvements

- Deploy the application to a cloud platform.
- Implement authentication and authorization.
- Add monitoring and logging tools.
- Scale the application using Kubernetes.

- **Author:** Cristhian Guerrero
- **Email:** [cristhian3815@gmail.com](mailto:cristhian3815@gmail.com)
- **LinkedIn:** [linkedin.com/in/cristhian-guerrero](https://linkedin.com/in/cristhian-guerrero)
- **Title:** MLOps Computer Engineer