MLOps Project Summary

End-to-End MLOps Pipeline for California Housing Regression

Objective

Build a reproducible, automated, and observable MLOps pipeline that trains, tracks, packages, deploys, and monitors a machine learning model using the California Housing dataset.

High-Level Architecture

1. Data Layer

- Raw Data: Fetched using fetch_data.py and stored in data/raw/
- Processed Data: Cleaned and split using preprocess.py into data/processed/
- New Data: Dropped into data/new/ to trigger retraining

2. Model Training & Tracking

- Training Scripts: train_linear.py, train_tree.py
- Experiment Tracking: MLflow logs metrics, parameters, and artifacts
- Model Selection: select best and register.py picks best model and registers it

3. API Layer

- FastAPI App: Serves predictions via /predict endpoint
- Swagger & ReDoc: Auto-generated interactive docs
- Health & Metrics: /health and /metrics endpoints for observability

4. Automation & Deployment

- Docker Compose: Orchestrates services locally
- **GitHub Actions**: CI/CD pipeline for build, test, and deploy
- **DVC**: Tracks and versions raw data

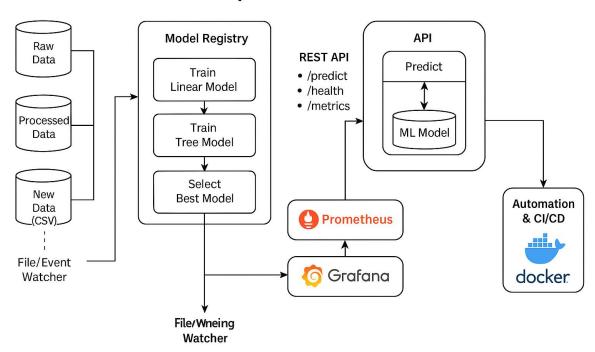
5. Monitoring & Observability

- Prometheus: Scrapes metrics from FastAPI
- **Grafana**: Visualizes request counts, latency, error rates
- Custom Dashboards: FastAPI observability dashboard (ID: 16110)

6. Retraining Trigger

- File Watcher: Monitors data/new/ using PollingObserver
- Trigger Logic: On new .csv, runs full pipeline via run training pipeline.py

MLOps architecture



Tools & Technologies

Category	Tools Used		
January 1	100.000		
Language	Python 3.8+		
API Framework	FastAPI		
Experiment Tracking	MLflow		
Containerization	Docker, Docker Compose		
CI/CD	GitHub Actions		
Monitoring	Prometheus, Grafana		
Data Versioning	DVC		
Logging	Custom logger via utils/logger.py		

Key URLs (Local Deployment)

Service	LIRI		
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FastAPI	http://localhost:8000
Swagger UI	http://localhost:8000/docs
MLflow UI	http://localhost:5000
Prometheus	http://localhost:9090
Grafana	http://localhost:3000

Outcomes

- Modular, reproducible ML pipeline
- Automated retraining on new data
- Real-time monitoring and metrics
- Scalable deployment via Docker
- CI/CD integration for production readiness