🔍 California Housing Price Prediction – Full Stack ML Project Summary

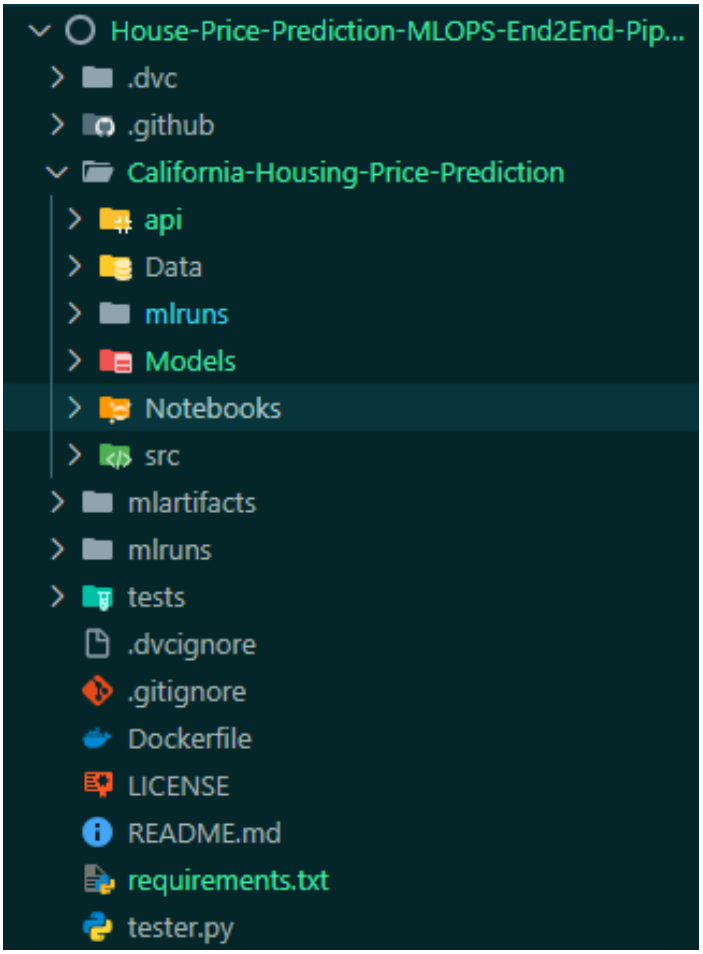
📁 Assignment Structure Overview

Project Details:-

GitHub Repo Link: - https://github.com/Atharv-Chaudhari-Bits/House-Price-Prediction-MLOPS-End2End-Pipeline

Docker Hub Link: - https://hub.docker.com/r/atharvchaudharibits/california-housing-app

Directory Structure –



Built and deployed an end-to-end California Housing Price Prediction pipeline using MLOps best practices. Implemented model training, tracking, and versioning with Git, DVC, and MLflow. Packaged the model as a REST API using FastAPI and containerized it with Docker. Automated CI/CD via GitHub Actions and set up basic logging with optional monitoring support.

✨ ML Models Implemented

We trained and compared several regression models using scikit-learn:

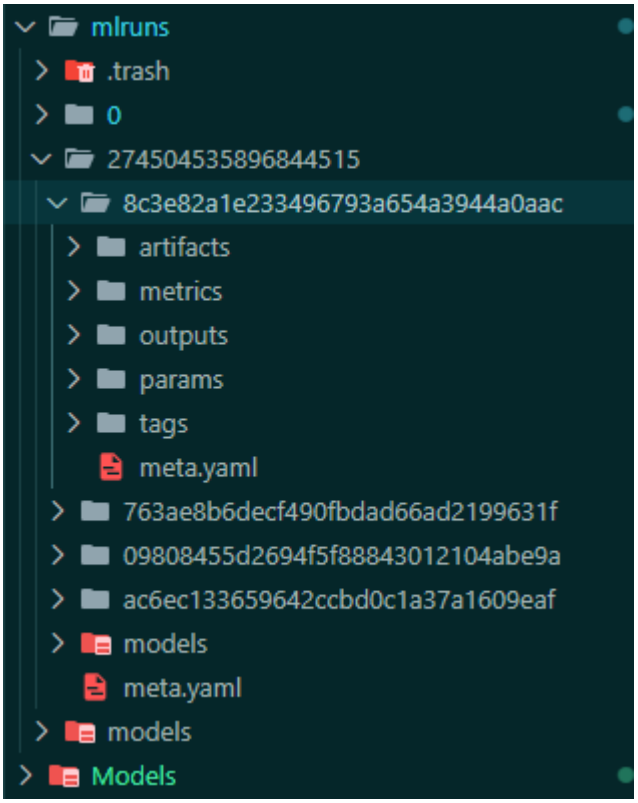
- `LinearRegression`
- `DecisionTreeRegressor`
- `RandomForestRegressor`
- `GradientBoostingRegressor`

Each model was evaluated and logged using **MLflow**, capturing:

- Hyperparameters
- Metrics (R^2, RMSE, MAE)
- Model artifacts
- Training metadata

MLflow created separate directories under `mlruns/<experiment-id>/<run-id>/`, which include:

- `params/`: model hyperparameters
- `metrics/`: performance metrics
- `artifacts/`: saved models
- `tags/`: metadata like model name or author



🚢 Docker Setup (Model Packaging + API)

The project is fully containerized using **Docker**, enabling smooth deployment across environments.

Key files used:

- `Dockerfile`: Defines the build, installing FastAPI, MLflow, scikit-learn, etc.
- `.dockerignore`: Ignores unnecessary files like `__pycache__`, notebooks, and MLflow trash.
- Docker builds the image with API code and launches a FastAPI server.

Main functionalities containerized:

1. Load the best MLflow model from registry
2. Serve predictions via REST (JSON) and HTML form (`/form`)
3. Logging for inference time, status, and model used

🌐 FastAPI Features

FastAPI is used to expose the model through:

- `/predict`: accepts JSON inputs and returns prediction
- `/form`: accepts user inputs via HTML form and returns output
- `/retrain`: accepts new CSV uploads for dynamic model retraining

Pydantic schemas are used for request validation.
All logs (including prediction attempts) are enhanced with **emojis/symbols** for better visibility in both console and file logs.

Logs are saved to app_log.log file

🔁 CI/CD with GitHub Actions

In `.github/workflows/`, I've set up a GitHub Actions pipeline that:

- Installs dependencies
- Runs test suite

🗃️ Data Versioning with DVC

DVC tracks both raw and processed data (`Data/raw`, `Data/processed`) and integrates with Git:

- Ensures reproducibility by locking datasets to specific model versions
- DVC files (`.dvc/`) and `dvc.yaml` track preprocessing and training steps
- Supports pushing large data to remote (e.g., S3, GDrive)

🎁 MLflow Tracking and Model Management

MLflow tracks:

- Training experiments
- Model performance
- Parameters and artifacts
- Registered models under unique IDs

✅ Logging and Monitoring

Logging is implemented for:

- API requests
- Inference status
- Retraining results

Logs include timestamps, status symbols (✅, ❌), and are stored in both console and file format.

⚙️ Retraining Pipeline via API

The `/retrain` endpoint supports:

- Uploading a new CSV via HTML form
- Reading and preprocessing new data
- Retraining the best model
- Logging it to MLflow
- Saving the new model to registry

🧪 Testing

Setup for `tests/` directory includes:

- Unit tests for preprocessing and inference
- API integration tests using `pytest`
- Retraining test cases for model updates

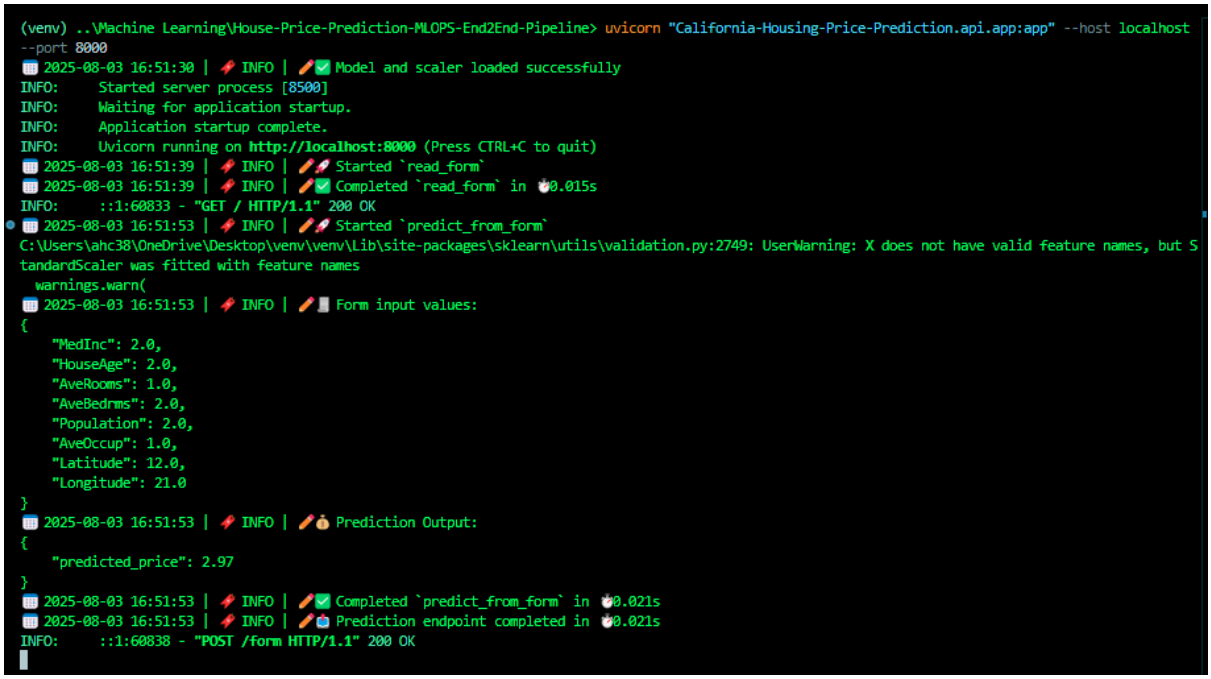
```
=> [internal] load build definition from Dockerfile                                                    0.2s
=> => transferring dockerfile: 536B                                                                    0.1s
=> [internal] load metadata for docker.io/library/python:3.12-slim                                     5.7s
=> [auth] library/python:pull token for registry-1.docker.io                                           0.0s
=> [internal] load .dockerignore                                                                       0.1s
=> => transferring context: 2B                                                                         0.0s
=> [1/5] FROM docker.io/library/python:3.12-slim@sha256:9c1d9ed7593f2552a4ea47362ec0d2ddf5923458a53d0c8e30edf8b398c94a31  37.5s
=> => resolve docker.io/library/python:3.12-slim@sha256:9c1d9ed7593f2552a4ea47362ec0d2ddf5923458a53d0c8e30edf8b398c94a31  0.1s
=> => sha256:f5cc5422ebcbbf01f9cd227d36de9dd7e133e1fc6d852f3b0c65260ab58f99f3 250B / 250B              0.8s
=> => sha256:4c665aba06d1c52829be84ca62e1030e27b8a3aa0f922666cbe74d24234ff227 3.51MB / 3.51MB          12.0s
=> => sha256:e3586b415667d044c3e5c7c91023d29d7db667b73a8082068a1b7f36c1962c34 13.66MB / 13.66MB        32.8s
=> => sha256:59e22667830bf04fb35e15ed9c70023e9d121719bb87f0db7f3159ee7c7e0b8d 28.23MB / 28.23MB        30.9s
=> => extracting sha256:59e22667830bf04fb35e15ed9c70023e9d121719bb87f0db7f3159ee7c7e0b8d               3.6s
=> => extracting sha256:4c665aba06d1c52829be84ca62e1030e27b8a3aa0f922666cbe74d24234ff227               0.3s
=> => extracting sha256:e3586b415667d044c3e5c7c91023d29d7db667b73a8082068a1b7f36c1962c34               1.3s
=> => extracting sha256:f5cc5422ebcbbf01f9cd227d36de9dd7e133e1fc6d852f3b0c65260ab58f99f3               0.0s
=> [internal] load build context                                                                       5.4s
=> => transferring context: 26.02MB                                                                    5.3s
=> [2/5] WORKDIR /app                                                                                  1.0s
=> [3/5] RUN apt-get update && apt-get install -y     gcc     && rm -rf /var/lib/apt/lists/*         115.4s
=> [4/5] COPY . .                                                                                      0.6s
=> [5/5] RUN pip install --no-cache-dir --upgrade pip     && pip install --no-cache-dir -r requirements.txt  1186.7s
=> exporting to image                                                                                  562.2s
=> => exporting layers                                                                                 348.4s
=> => exporting manifest sha256:c801cc2f578d60e73b46c4503aaae5d3e7bfc762d55b45966de70d4aaf6cf204       0.1s
=> => exporting config sha256:b32876d308b44640f35db7aee3e9705a481f8e52886474ab0d499506a14fe2c0         0.1s
=> => exporting attestation manifest sha256:050925223b611eea4687990be2fb6a64bfd606c00eec7df1fd64e713ab183b52  0.1s
=> => exporting manifest list sha256:5233fc527714e196fa5b0b1626eabffe35d753b9f62c54af952a90a05de38765  0.0s
=> => naming to docker.io/library/california-housing-app:latest                                        0.0s
=> => unpacking to docker.io/library/california-housing-app:latest                                     212.9s

View build details: docker-desktop://dashboard/build/desktop-linux/desktop-linux/qwusdbntwk1idc8lh2fnxpzeq
(venv) ..\Machine Learning\House-Price-Prediction-MLOPS-End2End-Pipeline>
```

train_test_split Aa ab .* No results

```python
import requests

url = "http://localhost:8000/predict"

payload = {
    "MedInc": 8.3252,
    "HouseAge": 41.0,
    "AveRooms": 6.9841,
    "AveBedrms": 1.0238,
    "Population": 322.0,
    "AveOccup": 2.5556,
    "Latitude": 37.88,
    "Longitude": -122.23
}

try:
    response = requests.post(url, json=payload)
    response.raise_for_status()
    print("✅ API responded successfully!")
    print("➡️ Input:", payload)
    print("📦 Response:", response.json())
except requests.exceptions.RequestException as e:
    print("❌ Error communicating with the API:")
    print(e)

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS

```
(venv) ..\Machine Learning\House-Price-Prediction-MLOPS-End2End-Pipeline> py tester.py
✅ API responded successfully!
➡️Input: {'MedInc': 8.3252, 'HouseAge': 41.0, 'AveRooms': 6.9841, 'AveBedrms': 1.0238, 'Population': 322.0, 'AveOccup': 2.5556, 'Latitude': 37.88, 'L
ongitude': -122.23}
📦 Response: {'predicted_price': 4.149}
(venv) ..\Machine Learning\House-Price-Prediction-MLOPS-End2End-Pipeline>
```



California House Price Predictor

Median Income: 2

House Age: 2

Average Rooms: 1

Average Bedrooms: 2

Population: 2

Average Occupancy: 1

Latitude: 12

Longitude: 21

Predict Price

Predicted House Price

$2.97k

Prediction made at: 2025-08-03 16:53:05

Metrics and Prometheus client Usage :-