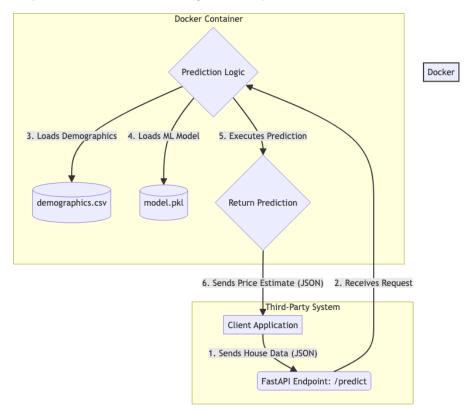
Technical Presentation: House Price Prediction API

1. Project Overview

Deployed a machine learning model for house price prediction as a scalable RESTful service. Key deliverables: API, test script, and model performance evaluation.

2. Architecture and Design Choices

Simple, robust architecture using modern open-source tools.



• API Framework: FastAPI

 $-\,$ High performance, asynchronous support, automatic validation, OpenAPI documentation.

• Containerization: Docker

Consistent environment for development/deployment, simplified dependency management.

• Data Handling:

- House data sent via JSON to /predict endpoint.
- zipcode_demographics.csv loaded into pandas DataFrame at startup.
- Production Consideration: Migrate demographic data to PostgreSQL for scalability.

3. Model Evaluation and Improvement

Initial Model

• Algorithm: KNeighborsRegressor with RobustScaler.

• Features: Small subset of numeric columns.

• Performance:

- **R-squared:** 0.7284

- Mean Absolute Error: \$102,337.19

Improved Model

Developed an improved model per project recommendations.

- Algorithm: GradientBoostingRegressor
 - More powerful ensemble method with higher accuracy than k-NN.
- Feature Engineering:
 - Used all available numeric features.
 - Extracted sale_year and sale_month from date field.
- Performance:
 - **R-squared: 0.8804** (+21% improvement)
 - Mean Absolute Error: \$69,841.29 (32% improvement)

Substantial improvement in prediction accuracy.

4. Scalability and Future Work

- Scalability: Containerized API allows easy scaling. Kubernetes deploys multiple instances with load balancing for high availability.
- Model Versioning: Current setup loads model at startup. Robust strategy:
 - 1. Store model artifacts in MLflow or S3.
 - 2. Create /-/reload_model endpoint for zero-downtime updates.
- Future Improvements:
 - Hyperparameter tuning with GridSearchCV.
 - Advanced feature engineering for geographical data.

5. Cloud Deployment Strategy

Production deployment in the cloud using:

Container Orchestration

- Kubernetes (preferred) or AWS ECS
 - Horizontal scaling based on demand
 - High availability through replica management
 - Rolling updates for zero-downtime deployments
 - Key Components: Deployments, Services, Ingress, ConfigMaps, Secrets

Cloud Provider Options

- AWS: EKS, ALB, S3, CloudWatch
- GCP: GKE, Cloud Load Balancing, Cloud Storage, Cloud Monitoring
- Azure: AKS, Application Gateway, Azure Blob Storage, Azure Monitor

Model Registry and Versioning

- MLflow (recommended):
 - Model registry with version control
 - Performance metrics comparison
 - Model stage transitions
- Alternatives: S3/GCS/Azure Blob with manual versioning, DVC, Weights & Biases

CI/CD Pipeline

- GitHub Actions or GitLab CI/CD:
 - Stages: Testing, Building, Security Scanning, Deployment
 - Tools: pytest, Docker Hub/ECR, kubectl, notifications

Monitoring and Observability

- APM: Prometheus, Grafana
- Metrics: Response times, throughput, error rates, resource utilization
- Logging: ELK Stack or similar solutions
- Health Checks: Liveness/readiness probes, downtime alerts

API Authentication and Security

- Methods: JWT, API Keys, OAuth2
- Implementation: FastAPI dependency injection, rate limiting
- Best Practices: TLS encryption, security scanning, least privilege

Auto-scaling

- Kubernetes Horizontal Pod Autoscaler (HPA)
- Scale based on CPU utilization or custom metrics (requests/second, response time)

Additional Considerations

- Backup/Recovery: Regular backups, cross-region replication
- Cost Optimization: Resource tuning, spot instances, auto-scaling policies

Comprehensive cloud deployment strategy ensures scalability, reliability, and maintainability.