## **Software Requirements Specification (SRS)**

Project Title: House Value Prediction System for California Real Estate Agency

Prepared by: Stefano Chen

Date: April 22, 2025

### 1. Introduction

### 1.1 Purpose

This document outlines the software requirements for a machine learning (ML) system designed to predict the market value of residential properties in California. Commissioned by a real estate agency, the goal is to enhance the accuracy and efficiency of property valuation processes.

#### 1.2 Scope

The system will provide property value predictions using a supervised ML model trained on historical property data. The system will expose a user-friendly interface and an API to allow agents to input property data and receive valuation predictions.

1.3 Definitions, Acronyms, and Abbreviations

• ML: Machine Learning

• API: Application Programming Interface

• MAE: Mean Absolute Error

• MAPE: Mean Absolute Percentage Error

• RMSE: Root Mean Square Error

# 2. Overall Description

### 2.1 Product Perspective

The system is a software solution that integrates with client data sources. It comprises:

- A backend ML engine for prediction
- A RESTful API for data communication
- A web-based frontend for internal users (real estate agents)

#### 2.2 Product Functions

- Accept property feature input and return estimated value.
- Train model on historical data with expert estimations.
- Retrain model with new data.
- Track prediction accuracy and provide performance metrics

#### 2.3 User Characteristics

- Real estate agents: Basic computer skills; use the interface to input data and receive predictions
- ML engineers: Responsible for model lifecycle management, retraining, and monitoring.

#### 2.4 Constraints

- The dataset contains expert estimations with a known average error of ~20%.
- Predictions must be returned within 2 seconds.
- Must comply with California data privacy laws.

#### 2.5 Assumptions and Dependencies

- Data provided is representative of the housing market.
- The system relies on stable internet connectivity for API and frontend access

## 3. Specific Requirements

#### 3.1 Functional Requirements

- FR1: The system shall be initially trained using the provided dataset.
- FR2: The system shall accept property feature inputs via the UI and API.
- FR3: The system shall return a predicted value with an associated confidence score.
- FR4: The system shall log all prediction inputs and results.
- FR5: The system shall support scheduled and event-based model retraining.

#### 3.2 Non-Functional Requirements

- **NFR1**: The system shall maintain a prediction error rate below 20%.
- **NFR2**: The system shall return prediction results within 2 seconds.
- NFR3: The system shall be deployed in a secure, scalable cloud environment.
- NFR4: The user interface shall be intuitive and accessible to non-technical users.
- NFR5: The system shall support version control for both models and data.

### 3.3 External Interface Requirements

- User Interface: A responsive web interface for data input and visualization.
- API Interface: RESTful endpoints for system integration and automation workflows.
- Data Interface: Input from CSV files and database connectors for model training and logging.

# 4. Appendices

### 4.1 Input Data Format

Column Name	Description
Median_House_Value	Median house value for household within a block (measured in USD) [prediction target]
Median_Income	Median income for households within a block of houses (measured in tens of thousands of USD) [10k\$]
Median_Age	Median age of a house within a block; a lower number is a newer building [years]
Tot_Rooms	Total number of rooms within a block
Tot_Bedrooms	Total number of bedrooms within a block
Population	Total number of people residing within a block
Households	Total number of households, a group of people residing within a home unit, for a block
Latitude	A measure of how far north a house is; a higher value is farther north [°]
Longitude	A measure of how far west a house is; a higher value is farther west [°]
Distance_to_coast	Distance to the nearest coast point [m]
Distance_to_LA	Distance to the centre of Los Angeles [m]
Distance_to_SanDiego	Distance to the centre of San Diego [m]
Distance_to_SanJose	Distance to the centre of San Jose [m]
Distance_to_SanFrancisco	Distance to the centre of San Francisco [m]

#### 4.2 Model Evaluation Metrics

- MAE (Mean Absolute Error): Measures average absolute difference between predicted and actual prices
- RMSE (Root Mean Square Error): Penalizes larger errors more heavily than MAE
- MAPE (Mean Absolute Percentage Error): Provides error in percentage terms for business interpretability