

Predicting Airbnb Listing Prices with MLflow and AWS S3

Project Overview

This project predicts optimal nightly prices for Airbnb listings on StayWise, a global vacation rental platform. Accurate price prediction helps hosts set competitive rates and improves booking efficiency. The project uses:

AWS S3: Data storage

MLflow: Experiment tracking and model registry

Machine Learning Models: Regression models for price prediction

Objectives:

Retrieve Airbnb listings from AWS S3

Clean and preprocess the data

Develop and compare multiple regression models

Track experiments and register the best model in MLflow.

Dataset Description

The dataset contains information about Airbnb listings, including price, location, amenities, reviews, and host details. The dataset requires preprocessing to handle missing values, categorical variables, and outliers.

Data Preprocessing:

Handled missing values in 'name' and 'host_name'.

Converted categorical variables using encoding.

Created new features like 'amenities_count'.

Removed outliers using the IQR method.

Split the dataset into training and testing sets.

MLflow Experiment Tracking:

All models were tracked using MLflow, logging metrics, parameters, and artifacts. The best model (Random Forest) was registered in the MLflow Model Registry. Attach MLflow UI screenshots here.

Repository Structure

```
airbnb-price-prediction/
  — notebook.py
  — requirements.txt
  — .gitignore
  — README.pdf
```

Setup Instructions:

1. Clone the repository

```
git clone https://github.com/yourusername/airbnb-price-prediction.git
cd airbnb-price-prediction
```

2. Install dependencies

```
pip install -r requirements.txt
```

3. Run Jupyter notebooks

```
jupyter notebook
```

Workflow Diagram:

1. Retrieve Data from S3
2. Data Cleaning & Preprocessing
3. Feature Engineering & Encoding
4. Train Regression Models
5. MLflow Experiment Tracking
6. Compare Models & Register Best
7. Deploy/Use Model

Data Preprocessing Steps

- Handle missing values (name, host_name, etc.)
- Encode categorical variables (neighbourhood, room_type)
- Handle outliers in price
- Feature engineering: reviews_per_month, amenities_count, etc.
- Normalize and scale numeric features

Model Development & MLflow

Tested regression models:

1. Linear Regression
2. Random Forest Regressor
3. Gradient Boosting Regressor

Metrics tracked using MLflow:

- RMSE
- R² Score

Model Performance

Model	RMSE	R ²
Linear Regression	45.56	0.54
Random Forest Regressor	43.88	0.57
Gradient Boosting	44.55	0.56

ML FLOW:

Experiment Tracking UI

The screenshot shows the 'Experiments' page of the MLflow UI. The left sidebar has 'Experiments' selected. The main area displays a table of experiments with columns: Name, Time created, Last modified, Description, and Tags. Two experiments are listed: 'testing' and 'Default'. The 'testing' experiment was created on 11/20/2025 at 07:34:40 PM and last modified on 11/20/2025 at 07:34:40 PM. The 'Default' experiment was created on 10/17/2025 at 09:08:14 PM and last modified on 10/17/2025 at 09:08:14 PM. There are buttons for 'Create', 'Compare', and 'Delete' at the top right.

Metrics Comparison:

The screenshot shows the 'Runs' page for the 'testing' experiment. The left sidebar shows 'Experiments > testing Machine learning'. The main area displays a table of runs with columns: Run Name, Created, Dataset, Duration, Source, and Models. Four runs are listed: 'Register_Best_Model' (Created 17 minutes ago), 'GradientBoosting' (Created 52 minutes ago), 'RandomForest' (Created 1 hour ago), and 'LinearRegression' (Created 1 hour ago). The 'Source' column shows paths like 'C:\Users...' and 'best_model'. The 'Models' column shows 'best_model', 'model', and 'model'. A link 'Show more columns (6 total)' is visible. At the bottom, it says '4 matching runs'.

mlflow 3.5.0

+ New

Home Experiments Models Prompts

Experiments > testing Machine learning

Runs Models Traces Share ...

Model attributes Model attributes

Model name	Status	Created	Logged from	Source run	Registered models	Dataset
best_model	Ready	20 minutes ago	C:\Users\aman2\AppData\Loca	Register_Best_Model	StayWisePriceModel v3	-
best_model	Ready	31 minutes ago	C:\Users\aman2\AppData\Loca	Register_Best_Model	StayWisePriceModel v2	-
model	Ready	54 minutes ago	C:\Users\aman2\AppData\Loca	GradientBoosting	-	-
model	Ready	55 minutes ago	C:\Users\aman2\AppData\Loca	RandomForest	-	-
model	Ready	1 hour ago	C:\Users\aman2\AppData\Loca	LinearRegression	-	-

3 8:39 2025-11-20

Overview Model metrics System metrics Traces Artifacts

Add tags

Model attributes

Type	Step	Model name	Status	Created	Registered models
Output	0	best_model	Ready	23 minutes ago	StayWisePriceM

Registered models

- StayWisePriceModel v3

Model Registres:

... About this run

Created at 11/20/2025, 08:18:47 PM
 Created by aman2
 Experiment ID 827181133532992984
 Status Finished
 Run ID 09e527b5aef94bf196826dc9f30c0ca6
 Duration 6.3s
 Source C:\Users\aman2\AppData\Local\ Packages\PythonSoftwareFoundation.Python.3.13_qbz5n2kfra8p0\Local Cache\local-packages\Python313\site-packages\ipykernel_launcher.py

Registered prompts —

Datasets None

Tags

8:40 PM 2025-11-20

Dependencies:

pandas
numpy
scikit-learn
mlflow
boto3
matplotlib
seaborn

Notes

Ensure AWS credentials have read access to S3.
Large datasets and MLflow artifacts are excluded via .gitignore.
Follow notebook execution order for smooth workflow.

Key insights and observations:

Features like location, number of amenities, and reviews significantly influence price.
Random Forest captures complex relationships better than Linear Regression or Gradient Boosting in this dataset.
Outlier removal improved model performance slightly.
MLflow provides a convenient way to compare model runs and manage model versions.