

A
Mini-project Report
on
“Housing Prices Predictor”

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INTRODUCTION

- It is imperative to find the correct price for a housing residence if anyone is planning to shift to a particular locality in a city with different brokers quoting different prices for similar houses in the same locality in a particular city.
- This project aims to reduce the burden of the common man who intends to find the probable range of prices for housing residences for a given parameters with minimal error.
- This project will encourage all the stake-holders in the real estate sector like brokers and the buyers as well as sellers to be more realistic when quoting housing prices and will give a base for understanding the market regarding the same.

Relavant Literature Survey

1. <https://github.com/supreethgowda/>

[House Price Prediction Project.git](#) : This is another project which predicts the Housing Prices with many different attributes like

- dayhours: Date house was sold
- price: Price is prediction target
- room_bed: Number of Bedrooms/House
- room_bath: Number of bathrooms/bedrooms
- living_measure: square footage of the home
- lot_measure: quare footage of the lot
- ceil: Total floors (levels) in house
- coast: House which has a view to a waterfront
- sight: Has been viewed
- condition: How good the condition is (Overall)
- quality: grade given to the housing unit, based on grading system
- ceil_measure: square footage of house apart from basement
- basement_measure: square footage of the basement
- yr_built: Built Year
- yr_renovated: Year when house was renovated
- zipcode: zip
- lat: Latitude coordinate
- long: Longitude coordinate
- living_measure15: Living room area in 2015(implies-- some renovations) This might or might not have affected the lotsize area
- lot_measure15: lotSize area in 2015(implies-- some renovations)

- furnished: Based on the quality of room 23: total_area: Measure of both living and lot

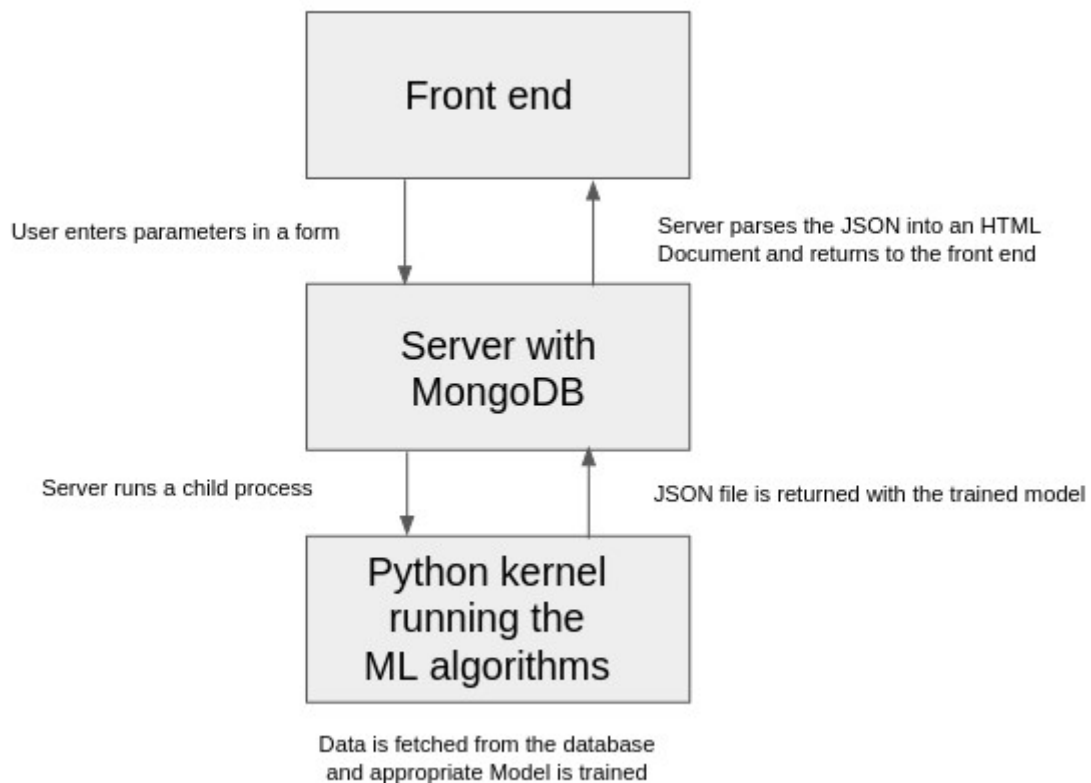
2. <https://towardsdatascience.com/machine-learning-project-predicting-boston-house-prices-with-regression-b4e47493633d> :

Machine Learning Project: Predicting Boston House Prices With Regression. In this project, we will develop and evaluate the performance and the predictive power of a model trained and tested on data collected from houses in Boston's suburbs.

List of Software/Hardware Requirements

In order to access the project and use it, one needs an internet connection, a device which can access Internet through web-browser, like PC or Mobile. We have made this project accessible as a web-apps, so general web-app dependencies apply like Browser version, etc. But no extra dependencies are required.

Proposed Architecture



- We have used HTML, CSS and Javascript for frontend. For Backend part, we have used Node.js and MongoDB.
- For editing the static pages we have used express which will make the pages dynamic.
- We have used MongoDB because it is a NOSQL database and data retrieval is faster.
- We have connected ML Model, MongoDB Database and run them with the help of Node.js.
- We run the Linear Regression from within a child process which is spawned from Node.js and runs python scripts and returns JSON files for the particular input given which are passed on to the front-end for viewing purpose.

Conclusion

- This project helped us understand the CRUD operations behind any database
- This project helped us to understand how to work in MongoDB and to connect MongoDB with Node.js using mongoose library.
- This project also helped us to understand linear regression in Machine Learning.

References

- GeeksForGeeks for Pymongo library.
- TutorialsPoint for basic MongoDB operations.
- GeeksforGeeks for Linear Regression.
- TowardsDataScience.com for data manipulation using Pandas library.
- MongoDB official documentation for predictive data models in MongoDB.

Annexure

AppsHTML / abbr — De...HTML Headings<h1>—<h6>: The H...color (CSS data typ...LoginWeb Development...Java | How to start L...Bitwise Operators ...PASC WebDev Cont...Reading list

Setup professional audio in "Audio Settings"

House Price Prediction

Area (Square Feet)

4000

BHK

12345
















































Bath

12345

Balcony

12345

Submit

entries							
	_id ObjectId	total_sqft Int32	bath Int32	balcony Int32	bhk Int32	price Mixed	
1	619f039799e7c14251466107	2850	4	1	4	428	  
2	619f039799e7c14251466108	1630	3	2	3	194	  
3	619f039799e7c14251466109	1875	2	3	3	235	  
4	619f039799e7c1425146610a	1200	2	0	3	130	  
5	619f039799e7c1425146610b	1235	2	2	2	148	  
6	619f039799e7c1425146610c	2750	4	0	4	413	  
7	619f039799e7c1425146610d	2450	4	2	4	368	  
8	619f039799e7c1425146610e	1875	3	1	3	167	  
9	619f039799e7c1425146610f	2065	4	1	3	210	  
10	619f039799e7c14251466110	2059	3	2	3	225	  
11	619f039799e7c14251466111	1394	2	1	2	100	  
12	619f039799e7c14251466112	1077	2	2	2	93	  
13	619f039799e7c14251466113	1566	2	0	2	180	  
14	619f039799e7c14251466114	840	2	2	1	50	  
15	619f039799e7c14251466115	1590	3	3	3	131	  
...							  


```

hello.py > ...
1  import pymongo
2  import sys
3  import pandas as pd
4  import numpy as np
5  from pymongo import MongoClient
6  client = MongoClient('localhost', 27017)
7  db = client.MINIPRO
8  collection = db.DATABASE
9  data = pd.DataFrame(list(collection.find()))
10 df=data[['total_sqft','bath','balcony','bhk','price']]
11 X = df.drop(df.columns[4],axis='columns')
12 Y=df[df.columns[4]]
13 from sklearn.model_selection import train_test_split
14 X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.2)
15 print(X_train)
16 from sklearn.linear_model import LinearRegression
17 lr = LinearRegression()
18 print(X_train.shape)
19 print(Y_train.shape)
20 Y_train=pd.DataFrame(Y_train)
21 lr.fit(X_train,Y_train)
22 l=[sys.argv[1],sys.argv[2],sys.argv[3],sys.argv[4]]
23 print(l)
24 print(lr.predict(sys.argv[1],sys.argv[2],sys.argv[3],sys.argv[4]))
25 l=[sys.argv[1],sys.argv[2],sys.argv[3],sys.argv[4]]
26 print(l)
27

```

```

const express = require("express");
const bodyParser = require("body-parser");
const mongoose=require("mongoose");
const app = express();

app.use(bodyParser.urlencoded({ extended: false }))
app.use(express.static("public"));
app.set('view engine', 'ejs');

mongoose.connect("mongodb://localhost:27017/miniproject")
const data=mongoose.Schema({
  square_foot:{
    type:Number,
  },
  bath:{
    type:Number,
  },
  balcony:{
    type:Number,
  },
  bhk:{
    type:Number,
  },price:{
    type:Number,
  }
})

const entry=mongoose.model("entry",data);
app.get("/",function(req,res)
{
  res.render("submit");
})
app.post("/",function(req,res)
{
  const sq_foot=req.body.Squareft;
  const bath=req.body.uiBathrooms;
  const balcony=req.body.uiBalcony;
  const bhk=req.body.uiBHK;

```