

# Web Application for House Price Prediction

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**Abstract** - This paper presents our work on an attempt to create a web application that can predict the price of a house using textual and visual features of the house. Our agenda was to produce a user-friendly software that is accessible to everyone. As building or buying a house is something every person does at some point in their lives, we thought this is a very essential and relevant project. To create the machine learning model two models were trained and merged in the end so that two different aspects of the house can be considered. Dense layers (model1) were used to train the textual features of the house which included number of bedrooms, number of bathrooms, area (in sq ft), and zipcode. For training the visual images that included the pictures of the bedroom, bathroom, kitchen and frontal elevation of the house CNN architecture was used (model2). Both model1 and model2 were combined to get the final model. As a frontend, a web application was created using streamlit where inputs (both textual and visual) can be inserted to get the predicted price of the house. Our research work was successful as we created a web application that can be used by everyone, even people who do not have any prior technological knowledge. The only prerequisite for using the web application is the requirement of data of the house

**Key Words:** Web application, house price prediction, Machine learning, Convolution Neural Network, Streamlit.

## 1. INTRODUCTION

Building a house can be a person's lifetime dream but a layman may not know the expenses for building the house and without proper planning and understanding they might have to spend a lot of money and sometimes even exhaust their savings. Traditionally estimating the building expenditure of the house is a tedious process. It is usually based on comparable sales price, the value of similar properties, and interference of unauthorized third parties, which is unreliable. It is always better to understand the expenses of building a house beforehand so that it can be adjusted according to one's likings. This prior knowledge of the expenditure of the house will help in efficiently managing one's finances and prevent repentance in the future.

This motivated us to pursue this research work to ensure that people are aware of the outlay. Precise automatic prediction for the houses prices is needed to help

policymakers to better design policies and control inflation and also help individuals for wise investment plans. Keeping in mind about all the people in the world we have created a web application that can be used by anyone with ease, even people without any technical knowledge.

## 2. RELATED WORK

To discover weaknesses in the current systems and improve its effectiveness, we must first comprehend it. The following is a list of current literature for the problem at hand.

The model[2] predicts the price based on the historical market price, present market value, and future development. It also features a user-friendly website where customers' requirements are collected. For data mining, the Naive Bayes method is employed. The disadvantage of this system is that they do not consider a visual data set.

The CNN model is used for feature extraction from satellite pictures[4,1], such as signs and text, and it predicts the price using linear regression on the limited data obtained after processing images. The following model makes use of a satellite to capture the street and aerial view, as well as the gathered dwelling qualities, such as location accessibility, neighborhood amenity, and structural elements. However, because a precise image of the house is not provided, the anticipated value may be erroneous. Instead, this model's predictions are made based on the area.

Another model[3] uses both texts and images, it is a comparative study of the SVM and simple Neural network model. Since the texts and images are not separated, the accuracy may be compromised.

There is, however, no system that offers a user-friendly online web app that uses both images and textual details of the house to predict the price, which is planned to be built.

## 3. PROPOSED SYSTEM

Dataset is the primary requirement of a machine learning model. Dataset for house prices that contains both visual and textual information composed of 535 sample houses from California state in the United States of America(U.S.A) is used here. Each house is represented by four images: a bathroom, a bedroom, a kitchen and a frontal view of the house. There are 2140 photos in the dataset folder. It also includes a text file with textual

information. The number of houses in order is represented by each row in the file. The textual data represents the house's physical qualities, such as the number of bedrooms, bathrooms, and square footage area, as well as the zip code of the location.

The main aim of our research is to build a user-friendly web application where the price of a house can be predicted using details including visual and textual features of the house. The proposed system's model was built using dense layers and convolution neural networks. Preprocessing of the dataset is done before building the model. Once all the preprocessing steps are done, the data is trained using a Multilayer Perceptron model that has been built using the Keras Functional API. A dense model is used to extract features from text data. Textual data is passed through dense layers and saved as an input model1. The CNN model is used to extract features from image data and saved as an input model2. The input model1 and input model2 are further merged and again passed through dense layers and finally through the flattening layer. Training of the model is stopped using the early stopping callback method and the best model weights are saved and exported.

The convolution layer is the core building block of CNN. This layer performs a dot product between two matrices, where one matrix is the set of learnable parameters, otherwise known as a kernel, and the other matrix is the restricted portion of the receptive field. The kernel is spatially smaller than an image but is more in-depth. This means that, if the image is composed of three (RGB) channels, the kernel height and width will be spatially small, but the depth extends up to all three channels. The rectifier linear activation function or ReLU is a piecewise linear function that will output the input directly if it is positive, otherwise, it will output zero.

The dense layer is a neural network layer that is connected deeply, which means each neuron in the dense layer receives input from all neurons of its previous layer. The dense layer is found to be the most commonly used layer in the models. The dense layer performs matrix-vector multiplication. The values used in the matrix are parameters that can be trained and updated with the help of backpropagation. The output generated by the dense layer is an 'm' dimensional vector. Thus, a dense layer is basically used to change the dimensions of the vector[8]. Dense layers also apply operations like rotation, scaling, and translation to the vector.

Once the model is built, the front end for the web app is created using the streamlit library. The web app contains different pages: one for the logo, a signup and login page, and the last page for entering the details of the house and getting the predictions. Each of these pages was designed separately and connected with one another using hyperlinks. The web app was then deployed using

streamlit sharing to obtain a public URL that can be accessed by anyone who has the link. This way, any person can get the price of the house without professional help or a third party.

#### 4. SYSTEM ARCHITECTURE

As shown in figure-1 our research includes two main components. First the web app for the user and second the model built by us. When a user launches the web app, he or she is presented with a login/signup screen. If they are a new user they are required to sign in by creating an account, or else they can just log in by entering their username and password. They are then directed to the home page where they should enter the input data required and once they click on the predict button they obtain the prediction stating the cost of the house.

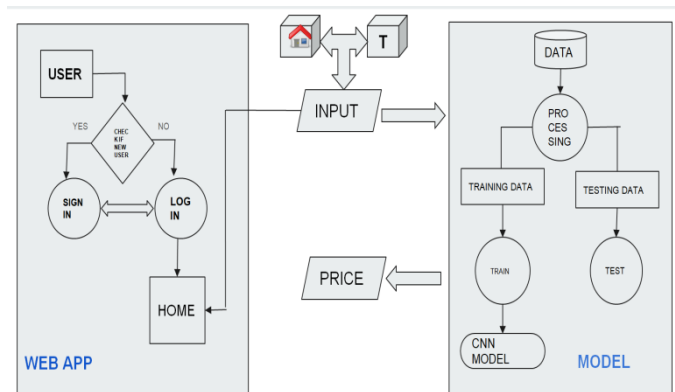


Fig -1: System Architecture

The data is preprocessed before building the model using various techniques considering each attribute. The preprocessed data is divided into two halves of training and testing data in a 70:30 ratio. Both the data sets are trained and tested to obtain satisfactory results.

The architecture of the model is as shown in figure-2. Textual data is passed through four dense layers and this is how input model 1 is formed. Image data is passed through convolution layers (Conv2D) twice and max-pooled each time. After flattening input model 2 is obtained. Both input model1 and model 2 are concatenated and passed through three more dense layers to obtain the final model of our research work to predict the price of the house.

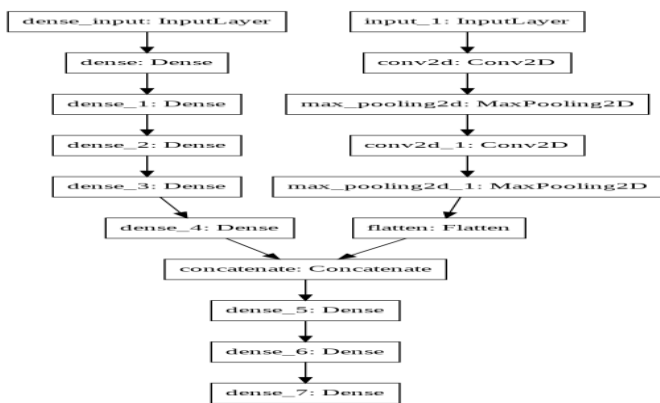


Fig -2: Model Architecture

## 5. EXPERIMENTS AND RESULTS

### 5.1. MODELS

As the dataset contain textual and image data both are preprocessed separately, outliers from textual data are removed. Normalization is performed on categorical data and the area is encoded as a Z score. Image data is resized and images corresponding to each house are concatenated. As the dataset contains both textual and image data both data is trained separately and combined. Textual data is trained using dense layers and images are trained using Convolution Neural Network.

### 5.2. REGRESSION CHART

Accuracy of the model can be obtained by plotting a regression chart of expected value and predicted value after testing.

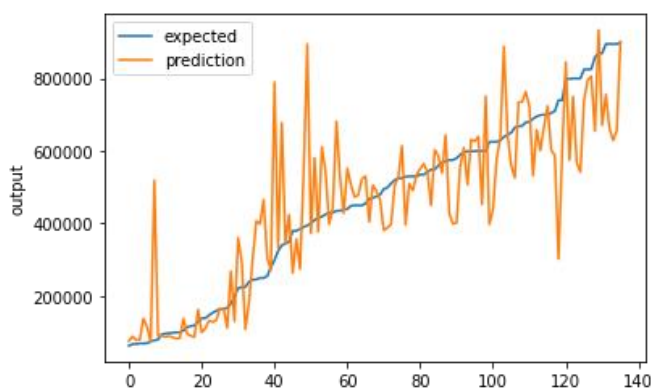


Fig -3: Regression Chart

The accuracy of the model is determined using the R2 score and RMSE value. R2 score is the goodness of fit of a regression model is represented by R-squared, it is a statistical measure.

$$R^2 = 1 - \frac{\sum(Y_i - \hat{Y})^2}{\sum(Y_i - \bar{Y})^2}$$

Where,

$\hat{Y}$ - predicted value of y

$\bar{Y}$ - mean value of y

The square root of the mean of the squared discrepancies between actual outcomes and predictions is used to calculate the RMSE.

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (predicted_i - actual_i)^2}{n}}$$

### 5.3. WEB APPLICATION

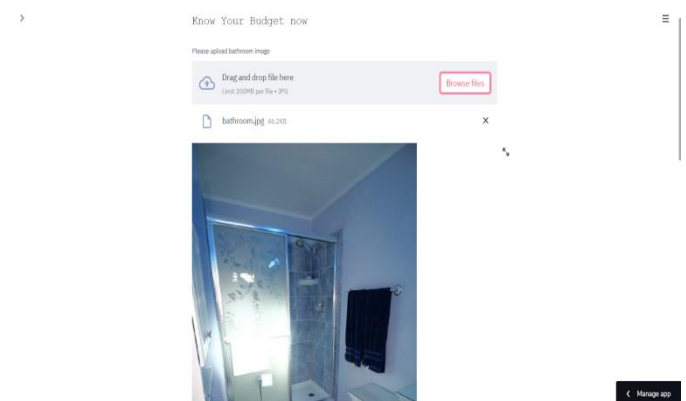
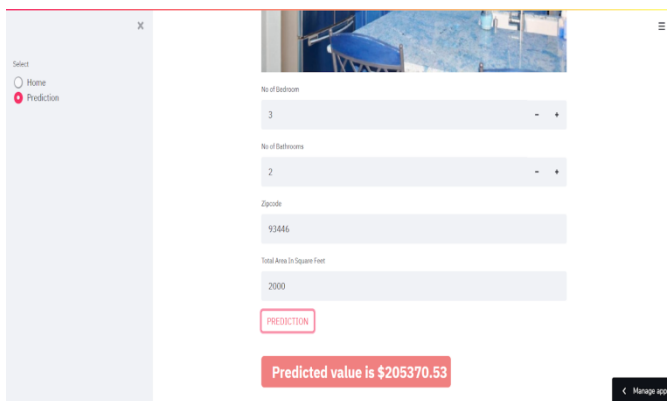


Fig -4: Uploading Image (Web App)

The final output of our work is a web application that was created using streamlit library. Streamlit is an open source python library that is used to make web applications according to the preference and liking of the creator. We have used Cascading Style Sheets (CSS) along with streamlit for coloring and styling the web app.

As shown in figure 4 and figure 5 the images of the house can be inserted along with the textual details as asked in the web application, in order to get the predicted price of the house. The web application was deployed using streamlit sharing and so the URL for the application was obtained.



**Fig -5:** Entering textual inputs (Web app)

- [7] <https://www.kaggle.com/amir22010/house-price-estimation-from-image-and-text-feature/data>
- [8] <https://machinelearningknowledge.ai/keras-dense-layer-explained-for-beginners/>

## 6. CONCLUSION

The objective was to create a system that would assist individuals in managing their finances in order to build their dream house. The ultimate result is a user-friendly web application that assists any group of people in estimating the value of their house by uploading images and textual data. The CNN assisted in the development of an accurate model. This is the first paper, to our knowledge, to have provided a web application for the purpose of predicting the price of the house. This may also be utilized in real estate and for other applications. In the future we can alter the dataset to improve the model by adding more zipcodes and other attributes.

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- [5] <https://streamlit.io/>
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