



A Preliminary Project Report on

SMART REAL ESTATE ASSESSMENT

Presented By

Patne Nikhil
Shinde Abhishek
Gholap Satyawan
Dange Namrata
Beera Vivechana



DEPARTMENT OF COMPUTER ENGINEERING
SVPML'S COLLEGE OF ENGINEERING
MALEGAON(Bk)
2018-19

Sponsored by



Xento Systems Private Limited, Pune

SMART REAL ESTATE ASSESSMENT

GITHUB.COM/ABHIESHINDE/SMART-REAL-ESTATE-ASSESSMENT

This research was done under the guidance of Prof. Nimbalkar S.S. with in the sponsorship of Xento Systems Private Limited, Pune, from August 2018.



SVPM's COLLEGE OF ENGINEERING DEPARTMENT OF COMPUTER ENGINEERING

CERTIFICATE

This is to certify that the Project Entitled

Smart Real Estate Assessment

Submitted by

Dange Namrata Milind
Gholap Satywan Ashok
Patne Nikhil Sunil
Shinde Abhishek Sunil
Beera Vivechana Ramesh

Exam No:B150194**206**
Exam No:B150194**214**
Exam No:B150194**234**
Exam No:B150194**242**
Exam No:B150194**259**

is a bonafide work carried out by Students under the supervision of
Prof. Nimbalkar S.S. and it is submitted towards the partial fulfillment of the requirement of Bachelor of Engineering (Computer Engineering) Project.

Prof.Nimbalkar S.S
Internal Guide

Prof.Kumbhar H.R
H.O.D

External Examiner

Principal

Place : SVPM's COE Malegaon(Bk.)
Date :

Acknowledgments

*It gives us great pleasure in presenting the preliminary project report on '**Smart Real Estate Assessment**'.*

*I would like to take this opportunity to thank my internal guide **Prof. Nimbalkar S.S** and all teachers for giving me all the help and guidance I needed. I am really grateful to them for their kind support. I appreciate all valuable suggestions given and queries raised by her which really helped us in exploring project in deep.*

*I am also grateful to **Prof.Kumbhar H.R**, Head of Computer Engineering Department, SVPM COE for his indispensable support, suggestions.*

*In the end our special thanks to **Mr.Kadam S.P** for providing various resources such as laboratory with all needed software platforms, continuous Internet connection, for Our Project.*

Dange Namrata
Gholap Satywan
Patne Nikhil
Shinde Abhishek
Beera Vivechana

(B.E. Computer Engg.)

Smart Real Estate Assessment

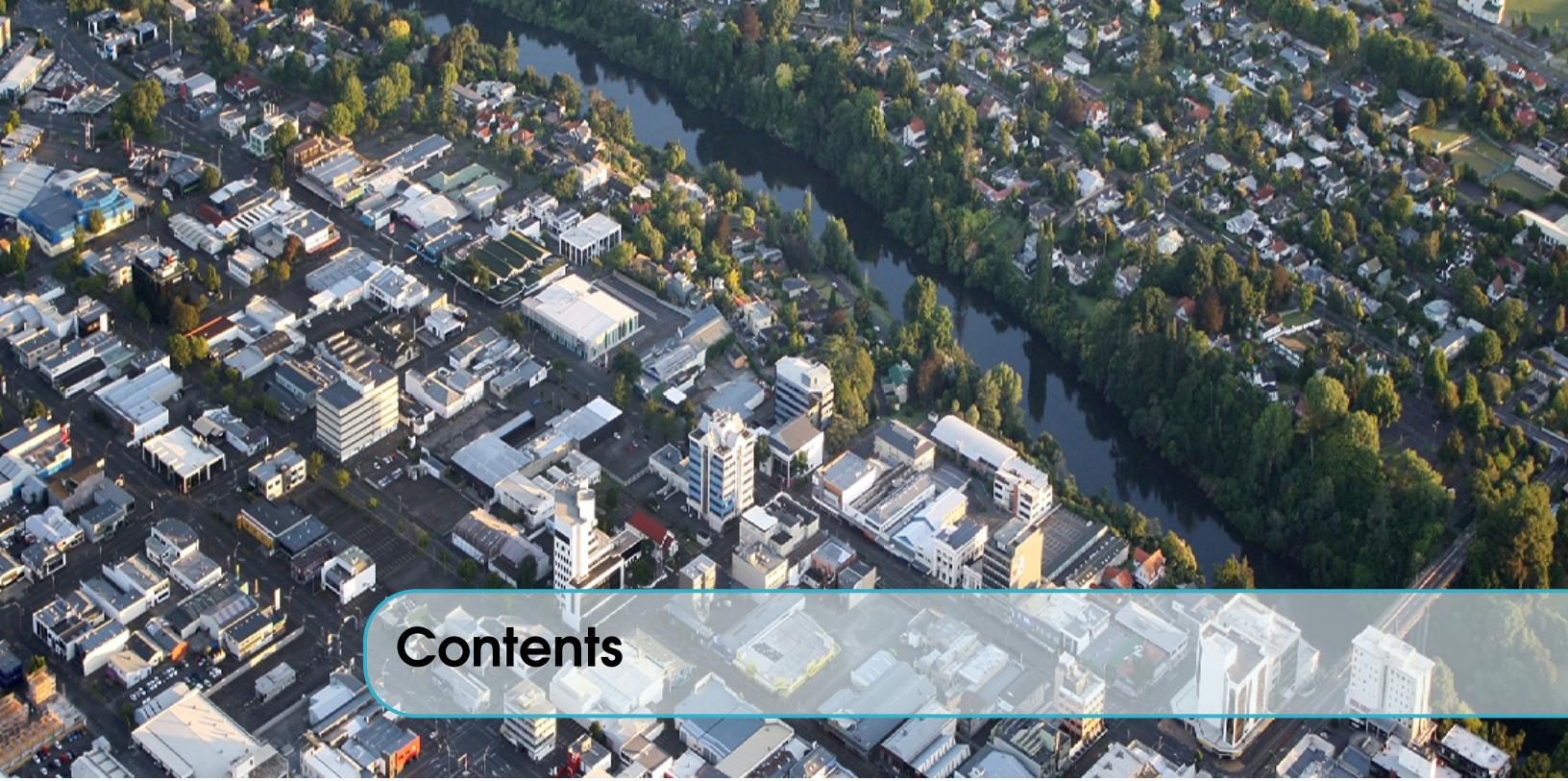
December 20, 2018

Abstract

REAL estate appraisal, which is the process of estimating the price for real estate properties, is crucial for both buys and sellers as the basis for negotiation and transaction. Traditionally, the repeat sales model has been widely adopted to estimate real estate price. However, it depends the design and calculation of a complex economic related index, which is challenging to estimate accurately. Today, real estate brokers provide easy access to detailed online information on real estate properties to their clients. We are interested in estimating the real estate price from these large amounts of easily accessed data. In particular, we analyze the prediction power of online house pictures, which is one of the key factors for online users to make a potential visiting decision. The development of robust computer vision algorithms makes the analysis of visual content possible. In this work, we employ a Recurrent Neural Network (RNN) to predict real estate price using the state-of-the-art visual features.

In a smart city, effective and accurate real estate assessments governed by a local government is crucial for determining the property taxes. Such assessments have never been trivial, and inappropriate assessments may result in disputes between property owners and the local government.

Generally for price prediction **Regression** is used (Prediction of continuous valued-function). But here we are going to use **Structured Deep Neural Network** in order to improve efficiency and accuracy. We introduce a deep learning approach to smartly and effectively assessing real estate values. We propose a systematic method to derive a layered knowledge graph and design a structured Deep Neural Network (DNN) based on it. Neurons in a structured DNN are structurally connected, which makes the network time and space efficient; and thus, it requires fewer data points for training. The structured DNN model has been designed to learn from the most recently captured data points; therefore, it allows the model to adapt to the latest market trends.



Contents

Acknowledgment	2
Abstract	2
List of Figures	ii
List of Tables	iii
1 Synopsis	v
 1.1 Project Title	v
 1.2 Project Option	v
 1.3 Internal Guide	v
 1.4 Sponsorship and External Guide	v
 1.5 Problem Statement	vi
 1.6 Goals and Objectives	vi
1.6.1 Goal	vi
1.6.2 Objectives	vi
2 Technical Keywords	vii
 2.1 Area of Project	vii
 2.2 Technical Keywords	vii

3	Introduction	ix
3.1	Project Idea	ix
3.2	Motivation of the Project	ix
3.3	Literature Survey	x
4	Problem Definition and Scope	xiii
4.1	Problem Statement	xiii
4.1.1	Goals and objectives	xiii
4.2	Major Constraints	xiii
4.3	Outcome	xiv
4.4	Applications	xiv
4.5	Hardware Resources Required	xiv
4.6	Software Resources Required	xv
4.6.1	Developing Environment	xv
4.6.2	Operating Environment	xv
5	Project Plan	xvii
5.1	Estimates	xvii
6	Algorithms	xix
6.1	Algorithm 1	xix
6.2	Algorithm 2	xx
7	Software Requirement and Specification	xxiii
7.1	Usage Scenario	xxiii
7.1.1	User profiles	xxiii
7.1.2	Usecase Diagram	xxiv
7.2	Functional Model and Description	xxiv
7.2.1	Data Flow Diagram	xxiv
7.2.2	Activity Diagram	xxvii
7.2.3	Non Functional Requirements:	xxviii
7.2.4	State Diagram:	xxix
7.2.5	Software Interface Description	xxx

8	Detailed Design Document	xxxii
8.1	Introduction	xxxii
8.2	Architectural Design	xxxii
8.2.1	Data Collectin and Preprocessing	xxxii
8.2.2	Layered Knowledge Graph	xxxii
8.2.3	Structured DNN	xxxiii
8.3	Component Design	xxxiv
8.3.1	Class Diagram	xxxiv
8.3.2	Sequence Diagram	xxxv
8.3.3	Deployment Diagram	xxxvi
9	Conclusion and Future Scope	xxxvii
References		xxxix
About Us		xl



List of Figures

2.1 Difference between Simple Neural Network and Deep Neural Network, https://becominghuman.ai/deep-learning-made-easy-with-deep-cognition-403fbe445351	vii
5.1 Benefits of Agile Model	xvii
5.2 Agile Model	xviii
7.1 Level 0 Data Flow Diagram	xxv
7.2 Level 1 Data Flow Diagram	xxv
7.3 Level 2 Data Flow Diagram	xxvi
7.4 Activity Diagram	xxvii
7.5 State transition diagram	xxx
8.1 Architecture diagram	xxxii
8.2 Class Diagram	xxxiv

8.3	Sequence Diagram	xxxv
8.4	Deployment Diagram	xxxvi



List of Tables

4.1	Hardware Requirements	xiv
4.2	Hardware Requirements	xiv



1. Synopsis

1.1 Project Title

Smart Real Estate Assessment

1.2 Project Option

Internal project

1.3 Internal Guide

Prof.Nimbalkar S.S

1.4 Sponsorship and External Guide

Xento Systems Private Limited, Pune

Mr. Amol Deshpande

Mr. Mayur Narole

Mr. Anand Chandole

1.5 Problem Statement

To predict the price of the real estate properties which depends upon various factors like Infrastructure, Neighbourhood environment, House age, Location etc. using Structured Deep Neural Network.

1.6 Goals and Objectives

1.6.1 Goal

To construct an assessment system using Deep Neural Network in a more efficient effective way with, improved accuracy for house-price prediction.

1.6.2 Objectives

- Prediction of prices of real estate property more accurately(Using DNN).
- Elimination of Role of Middle Person(*Agent*).
- Assessments are not only for urban areas but also for Rular areas.
- Real time prediction of can be possible

2. Technical Keywords

2.1 Area of Project

Deep Neural Network

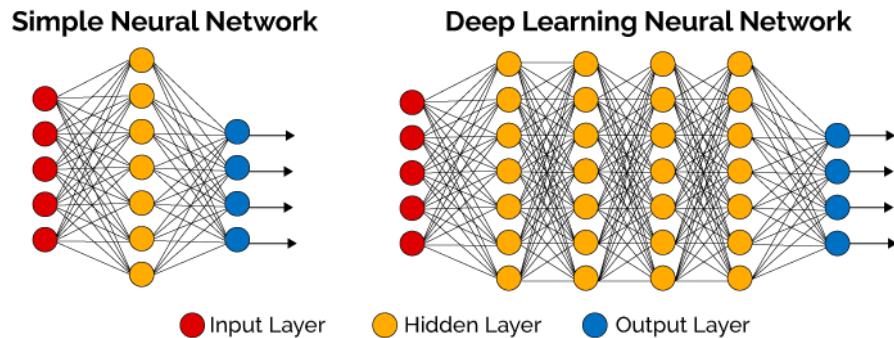


Figure 2.1: Difference between Simple Neural Network and Deep Neural Network,
<https://becominghuman.ai/deep-learning-made-easy-with-deep-cognition-403fbe445351>

2.2 Technical Keywords

1. **Deep Learning:** Deep learning is an aspect of artificial intelligence (AI) that is concerned with emulating the learning approach that human beings use to gain certain types of knowledge. At its simplest, deep learning can be thought of as a way to automate predictive analytics. Deep learning is the branch of machine learning which

having ability to learn multiple level representation from structured data using supervised or unsupervised learning.

2. **Personalization:** Personalization goes more with the user's decisions, this means, the instructions explicitly given. An example could be the background image of twitter could be taken as personalization of your twitter profile.
3. **Online Interest:** Get users interest through social media like twitter,facebook. For that data need to mine.
4. **Social Media:** To mine users point of interest need to get data from users social account.



3. Introduction

3.1 Project Idea

Real estate appraisal, which is the process of estimating the price for real estate properties.

A report published by EPRA(*Europian Public Real Estate Association*) real estate in all its forms accounts approximately 18-20 percent of its economical activity.therefore accurate prediction of real estete properties are crucial. For most of the working classes housing has been one of the largest expense, so to make right decision on the real estate investment is much cruicial.

3.2 Motivation of the Project

As we seen the report published by EPRA and why investment in housing is important the thats why accurate preduction of real estate properties are crucial.Traditionally previous work for prediction of price is based on regression analysis and machine learning.but due to rapid development in Deep Neural Network feild it is much benificial to use DNN instead of Regression.Because of recently developed deep learning,computer becomes smart enough to interpret visual content in similar way that human

can.

peoples can able to estimate price more accurately in this system as compared to previous systems only because of DNN.

3.3 Literature Survey

 J. Frew and G. Jud, "Estimating the Value of Apartment Buildings," Journal of Real Estate Research

- Techniques : Hedonic modeling techniques to estimate
- Advantages : Able to estimate prices correctly in proportion of size and number of units
- Disadvantages : Effect of Aging isn't considered effectively

 R. E. Lowrance, "Predicting the Market Value of Single-Family Residential Real Estate," Technical Report

- Techniques : Local Linear model and Random Forest model
- Advantages : lowest expected error on unseen data and model is tailored to zip codes using indicator variables
- Conclusion : Random forest model may perform better than the local linear model

 X. Hu and M. Zhong, "Applied Research on Real Estate Price Prediction by the Neural Network,"

- Techniques : Backpropagation neural networks and Elman neural network
- Advantages : Elman neural network could forecast more accurate and constringe faster than other approaches

 N. Nguyen and A. Cripps, "Predicting Housing Value: A Comparison of Multiple Regression Analysis and Artificial Neural Networks," Journal of Real Estate Research

- Techniques : ANN and multiple regression analysis
- Advantages : when enough data points were available for training, ANNs could perform better than multiple linear regressions

 Y. E. Hamzaoui and J. A. H. Perez, "Application of Artificial Neural Networks to Predict the Selling Price in the Real Estate Valuation Process"

- Techniques : Feed-forward backpropagation neural network with a single hidden layer
- Advantages : reliable prediction of house selling prices at that time

 X. Zhang, "Using Fuzzy Neural Network in Real Estate Prices Prediction"

- Techniques : fuzzy neural network to support fuzzy reasoning and learning
- Advantages : works better than traditional neural network approaches

R S. Chopra, T. Thampy, J. Leahy, A. Caplin, and Y. LeCun, “Discovering the Hidden Structure of House Prices with a Non-Parametric Latent Manifold Model”

- Techniques : Latent Manifold Model
- Advantages : with two trainable components like one parametric component that predicts the “intrinsic” price of a house using its features, and the second one is a non-parametric component that calculates the desirability of the neighborhood; it performs better than pure parametric or non-parametric models
- Disadvantages : Fails in Rular areas

R B. J. Ford, H. Xu, and I. Valova, “A Real-Time Self-Adaptive Classifier for Identifying Suspicious Bidders in Online Auctions”

- Techniques : Neural Network model
- Advantages : By training the neural network with newly added structured data, so it can quickly adapt to changing trends in bidding, model is able detect suspicious bidders in online auctions
- Disadvantages : used structured input and learning processes, and the DNNs are fully connected

R S. Zhang, Y. Bao, P. Zhou, H. Jiang, and L. Dai, “Improving Deep Neural Networks for LVCSR Using Dropout and Shrinking Structure”

- Techniques : Specialized structured neural network
- Advantages : Shrinking DNN structure with hidden layers decreasing in size from a lower layer to higher layers for the purpose of reducing the model size and making the model time efficient without affecting performance.
- Disadvantages : they failed to justify why shrinking DNN would not affect performance, and also there was a lack of systematic approach for the network reduction



4. Problem Definition and Scope

4.1 Problem Statement

To predict the price of the real estate properties which depends upon various factors like Infrastructure, Neighbourhood environment, House age, location etc. using *Structured Deep Neural Network*.

4.1.1 Goals and objectives

Goals

The Goal of this System is to predict the price of the real estate property more precisely

Objectives

- Prediction of prices of real estate property more accurately (Using DNN).
- Elimination of Role of Middle Person(Agent).
- Assessments are not only for urban areas but also for rural areas.

4.2 Major Constraints

- Application should be web based.
- Supported by any browser (mobile browser, pc browser).

- The information store will be an PostgreSQL database.

4.3 Outcome

- To get the estimated price of the real estate property.

4.4 Applications

- Housing Investment
- Auctions
- Real time Valuation

4.5 Hardware Resources Required

Developing Environment:

Sr. No.	Parameter	Minimum Requirement
1	CPU Speed	2.5 GHz
2	RAM	8 GB
2	HDD	1 TB

Table 4.1: Hardware Requirements

Operating Environment:

Sr. No.	Parameter	Minimum Requirement
1	CPU Speed	800 MHz
2	RAM	256 MB

Table 4.2: Hardware Requirements

4.6 Software Resources Required

4.6.1 Developing Environment

Platforms :

1. Operating System: Windows 10
2. IDEs: PHP Storm, Visual Studio Code.
3. Programming Language :PHP,Python
4. Database: PostgreSQL .

5. Tools:

- (a) Documentation: TexLive(Texworks Editor) and Overleaf(Online LaTeX Editor)
- (b) Diagram: LucidChart (Online Drawing Tool)

4.6.2 Operating Environment

Platforms :

1. Operating System:Windows/Linux Based version released after 2012,Mac
2. Web Browser: Chrome version released after 2012,Microsoft EDGE,Mozilla

5. Project Plan

5.1 Estimates

Agile model:

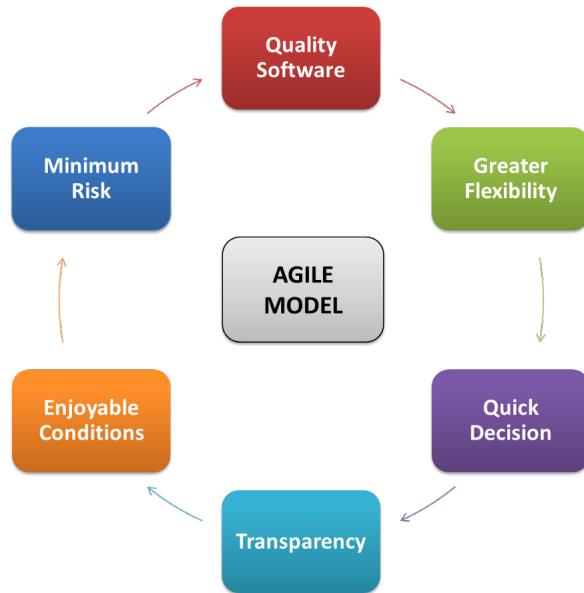


Figure 5.1: Benefits of Agile Model

©<https://medium.com/@proximitycrc/6-benefits-of-agile-model-bddf55f976b5>

The Agile Method is a particular approach to project management that is utilized in software development. This method assists teams in

responding to the unpredictability of constructing software. It uses incremental, iterative work sequences that are commonly known as sprints. Agile Methods break the product into small incremental builds. **Scrum** is a framework for Agile software development.

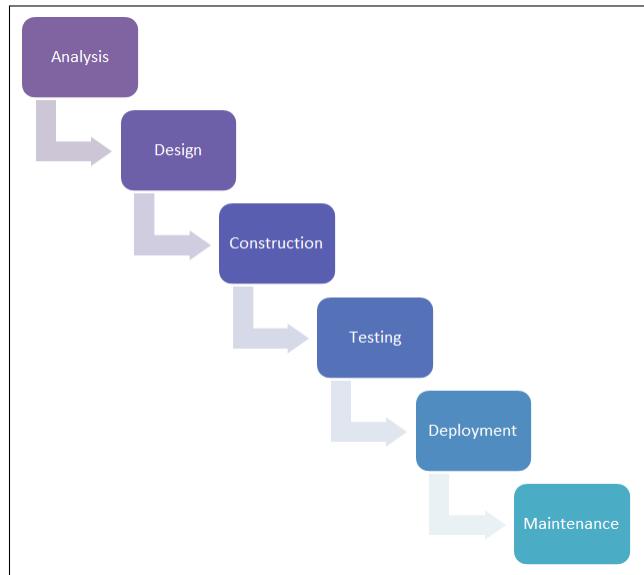


Figure 5.2: Agile Model

©<https://www.powerobjects.com/2013/02/11/microsoft-dynamics-crm-and-agile-development>

6. Algorithms

6.1 Algorithm 1

Name

Construct Structured Knowledge

Input

Training dataset and testing dataset in domain D

Output

Layered Knowledge Graph (G) for domain D

Description

In this algorithm we take the training and testing dataset as the input and construct a Knowledge Graph from it. Firstly in this we initialize the DNN which has k layers and $n(i)$ neurons in each layer. Then we train DNN using the training dataset which we have taken as input. And similarly we test the DNN using the testing dataset. At this point we record its accuracy. After training and testing the DNN, we identify the weak links, weak neurons and the weak layers .So basically the weak links are the weights are the links with weak weights that are below the predefined threshold. We delete these weak links from the DNN. Then we identify weak neurons; these are the neurons with very few links,

-
1. Initialize DNN Σ with k layers and n_i neurons in each layer.
 2. Train Σ using the training dataset.
 3. Test Σ using the test dataset and record its accuracy.
 4. Repeat the following until there is a significant decrease in accuracy
 5. Identify the weak links in Σ , and delete such links.
 6. Identify the weak neurons in Σ , and delete or combine them.
 7. Identify the weak layers in Σ , and delete or combine them.
 8. Train the reduced Σ using the training dataset.
 9. Test the reduced Σ using the test dataset and record its accuracy
 10. Restore Σ from its previous iteration.
 11. Name the hidden nodes and adjust Σ accordingly.
 12. Create Φ according to Σ by ignoring all weights.
 13. **return** layered knowledge graph Φ .
-

such links can also be deleted from the DNN. Or such neurons can be combined into a stronger one .After this we examine the hidden layers with very few neurons, which are called the weak layers. In a similar way we can delete the weak layer or combine two adjacent weak layers into a single one. After removal of all the outliers we again train the DNN using the training dataset and test the DNN using the testing dataset and we once again record the accuracy. We restore the resultant DNN which is a structured with no weak links or weak neurons or weak layers and we name the hidden nodes and adjust the DNN accordingly. Then we create a graph which is the layered Knowledge Graph.

6.2 Algorithm 2

Name

Create Structured DNN

Input

from algorithm 1, Layered Knowledge Graph G

Output

a structured DNN S for Domain D

1. Create Ψ with k layers and n_i neurons in each layer as in Φ .
 2. Let $\pi(n)$ be the mapping of node n in Φ to a neuron in Ψ
 2. **for** layer l from 1 to $k-1$ in Φ
 3. **for each** node n in layer l
 4. **if** node n has an edge to node x in layer $l+1$ of Φ
 Connect $\pi(n)$ and $\pi(x)$ in Ψ
 5. **for** layer l from 2 to k in Ψ
 6. **for each** neuron n in layer l
 7. add a bias node and connect it to neuron n
 8. **return** structured DNN Ψ
-

Description

A layered knowledge graph is used to derive structured DNN it is used for training and price prediction. Constructed DNN has 4 layers :Input layer,2 hidden layer, output layer.structured DNN trained using standard feedforward back-propagation algorithm.We initialize all connection weight with relatively small random value within range [-0.5,0.5].Structured DNN contains only needed connection between neurons,therefore it is Time and space efficient trained using fewer data points



7. Software Requirement and Specification

7.1 Usage Scenario

7.1.1 User profiles

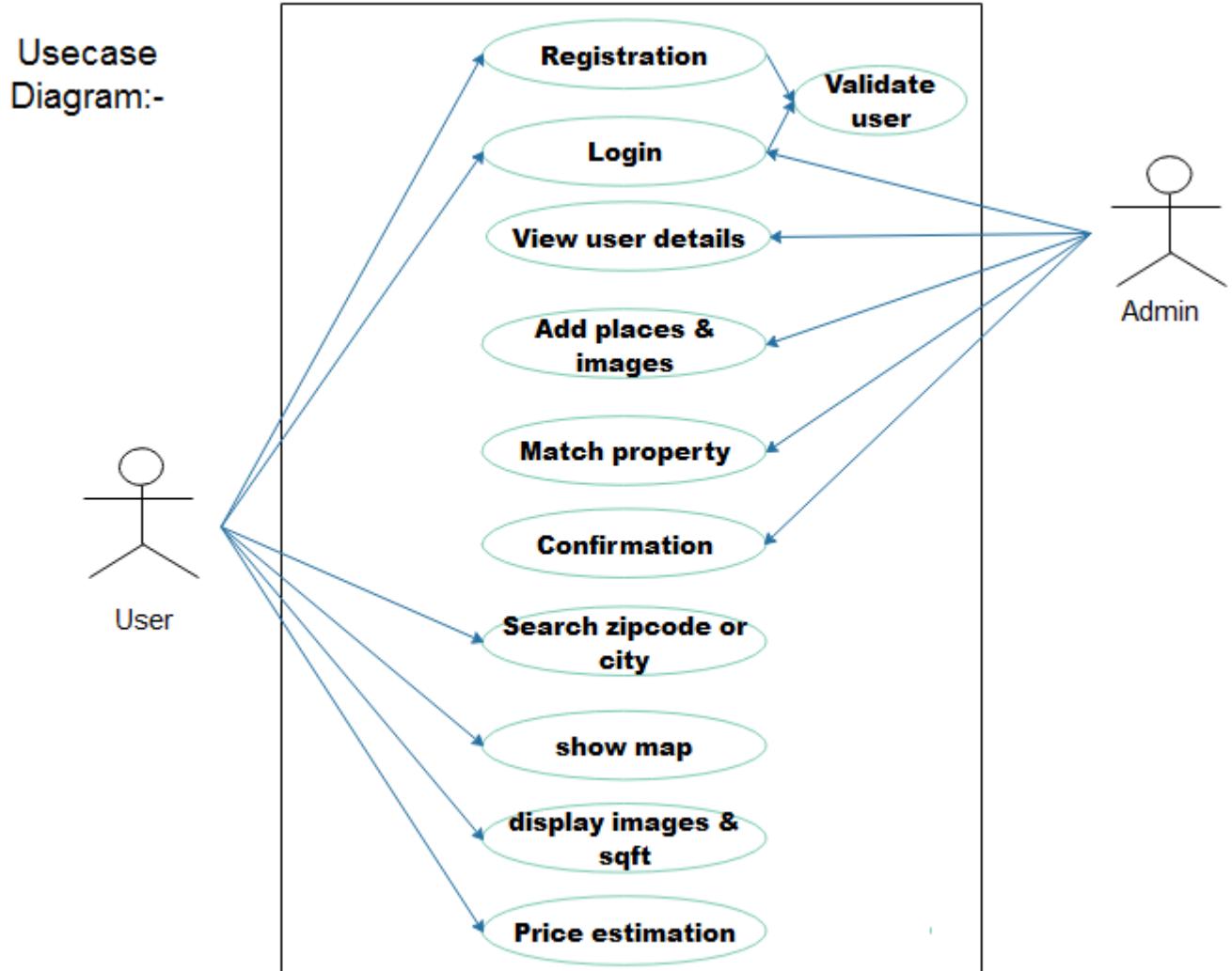
Admin

Admin manage the users and also set policies. It verifies the user and provides the access rights to the System. It is able to add new Data in the model.

User

User can be log in into the system and check for new places, properties and offers. They are able to add new property for sell, rent as well as for bidding.

7.1.2 Usecase Diagram



7.2 Functional Model and Description

7.2.1 Data Flow Diagram

Level 0 Data Flow Diagram

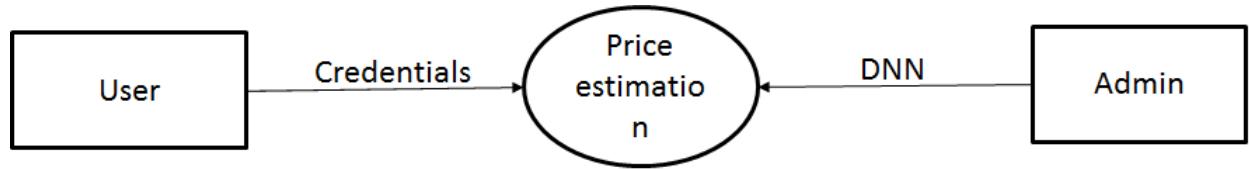


Figure 7.1: Level 0 Data Flow Diagram

Level 1 Data Flow Diagram

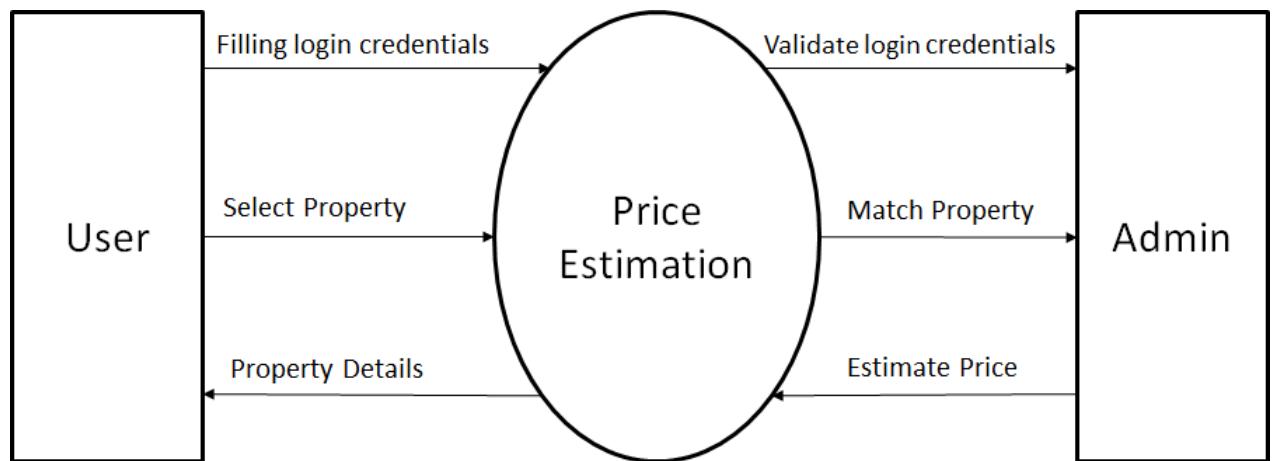


Figure 7.2: Level 1 Data Flow Diagram

Level 2 Data Flow Diagram

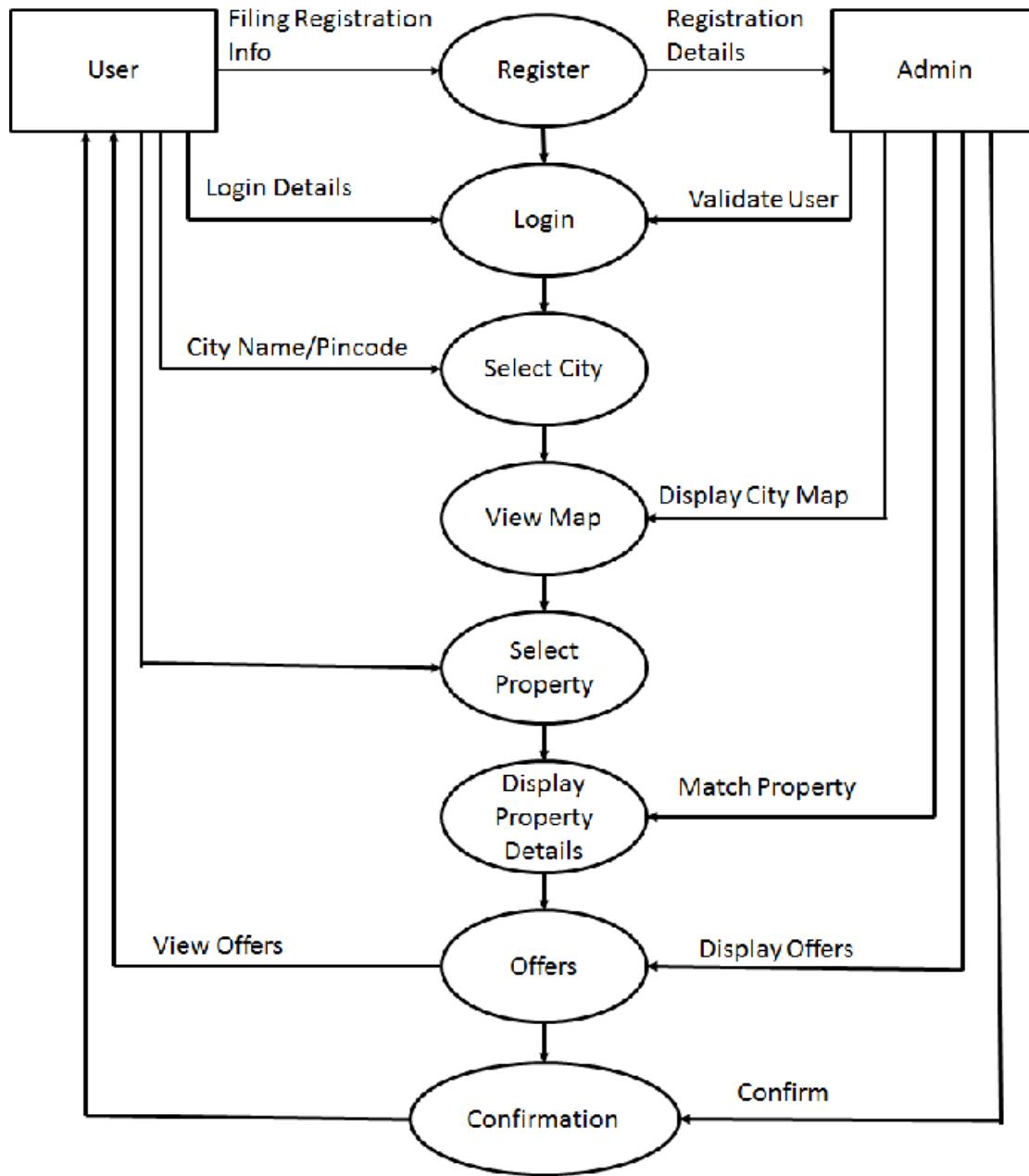


Figure 7.3: Level 2 Data Flow Diagram

7.2.2 Activity Diagram

The Activity diagram represents the steps taken.

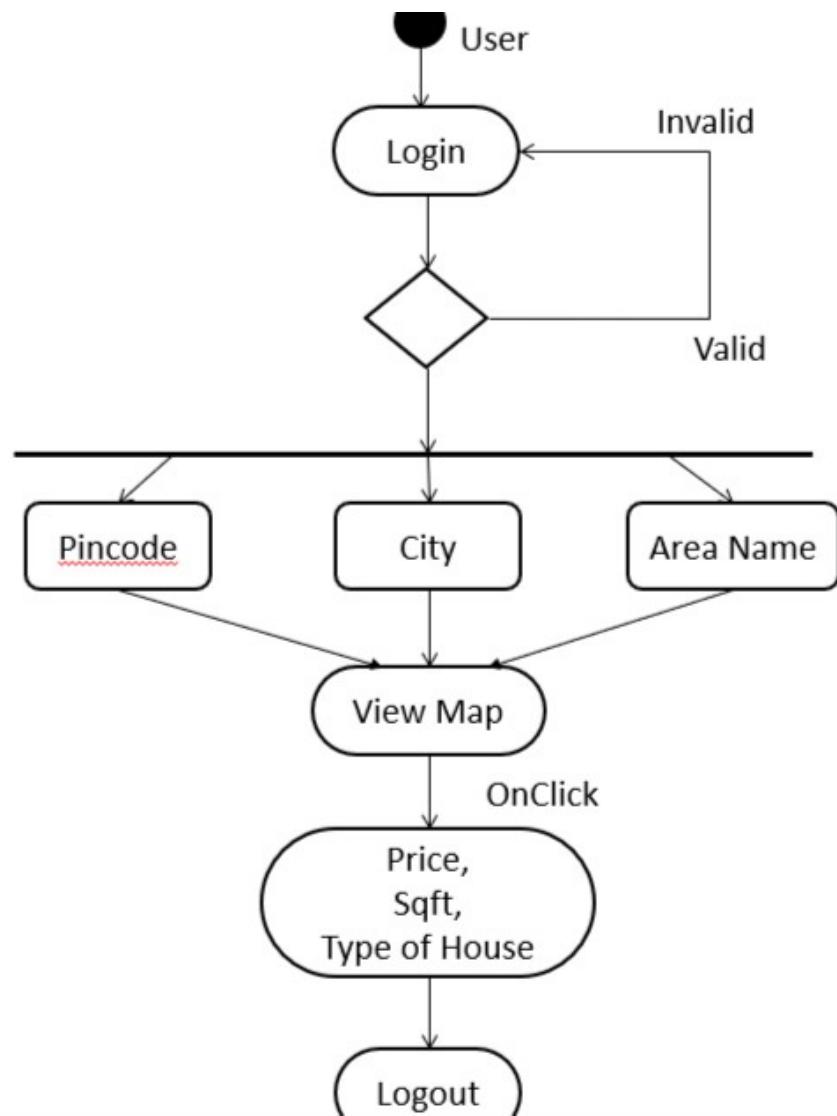


Figure 7.4: Activity Diagram

7.2.3 Non Functional Requirements:

Communication interface

It is a one-to-many user interface for communication. The users will use HTTP protocol to send queries to server. For this purpose HTML5 compatible browser is required and user can access the database through the browser by URL.

Performance Requirements

1. Performance should be high i.e. response time should be low in fraction of seconds.
2. Performance is increased by providing one URL to all clients through which they can access the database.
3. Proper updates of records database and some other information.
4. The system does not make any mistake.
5. Results needs to be shown to user in proper manner to make system effective.
6. Appropriate colour combinations should be used.
7. The GUI must be clear, unambiguous and exible for use.

Safety Requirements

As the database on server is centralized one there may be chances of data crash due to virus or other problems of software. So data backup is required. The record les are stored in another server and generated report files are stored on another server. Because of that no any conict occurs.

Security Requirements

Server Application is run only on the Server side, it provides Security and no chances of misuse. Proper authentication must be provided as all the data is on server. So even new admin registration must be provided only under senior admin permission . And only registered users can access the data and makes each other to their friends.

Software Quality Attributes

1. **Adaptability:** Any changes in software or any advanced modification will be possible. This project is reusable for creating new search

engines. The algorithms used in this project can be modified in future to get more effective performance in case of reducing time and space complexity.

2. **Maintainability:** Maintains performance throughout the system. As we are providing results to query instantly within fraction of seconds it improves performance of using this search engine greatly.
3. **Portability:** System is portable to any environment. Java ,ph,python are portable and platform independent language .So main aim of using the PHP for our project is to provide portability to our project.
4. **Reliability:** Having many qualities to make system reliable. We are providing backup and recovery facility in centralize environment. This makes our database more reliable.

Database Requirements

PostgreSQL Server connectors are required for the connectivity between PHP and PostgreSQL server database, and php and PostgreSQL connectivity is used to store and fetch data .

Internalization Requirements

To make our project global extensions required are different country languages e.g. English etc.

Legal Requirements

As our project is web based project multiple users may interact with this system simultaneously with different platforms on their PC. Also our project works according to concern of user expectation.

Reuse Objectives for the project

This project can be reused in future for developing new system which provides simultaneously access .Also battery efficiency will be done because of this will be used in future for sensing with least battery.

7.2.4 State Diagram:

Fig.7.5 example shows the **State Transition Diagram**. The states are represented in ovals and state of system gets changed when certain events

occur. The transitions from one state to the other are represented by arrows. The Figure below shows important states and events that occur while creating new project.

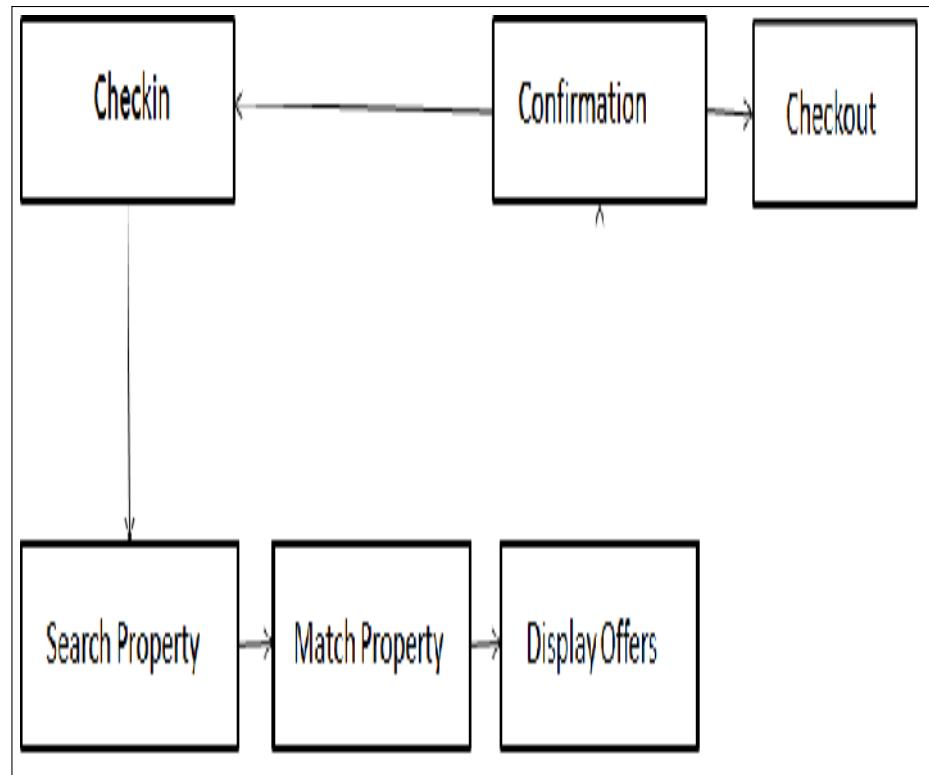


Figure 7.5: State transition diagram

7.2.5 Software Interface Description

Here we will be using internet connection and HTTP protocol for web based access



8. Detailed Design Document

8.1 Introduction

This document specifies the design that is used to solve the problem of Product.

8.2 Architectural Design

A description of the program architecture is presented. Subsystem design or Block diagram, Package Diagram, Deployment diagram with description is to be presented.

8.2.1 Data Collection and Preprocessing

We collected real estate data from a real estate website Zillow.com. This website conserves all recent and past house listings data including house features, market features, public records of houses, neighborhood features, etc. An entire number of 15 features are predefined and their associated values are collected. The predefined features include number of beds, number of baths, square footage, lot size, built year, yearly tax, similar houses average sold price, adjacent schools average ratings, fireplace, waterfront, the number of stories, heating, cooling, patio, and park. In

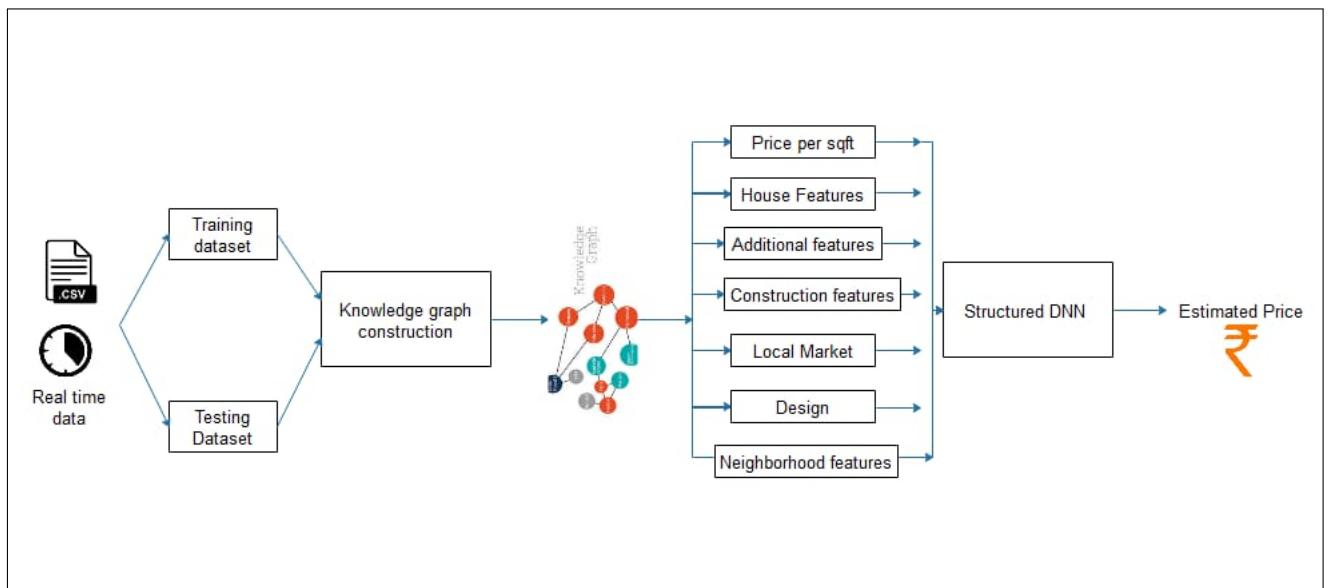


Figure 8.1: Architecture diagram

our approach, we first remove the following outliers: First is data points with insufficient information, e.g., missing square footage. And second is data points with unreasoning values such as a sold price of 1 as a gift. We additional calculate the average selling price of houses sold in last few months. Houses with too high or too low price-per-square feet are considered all outliers, and thus they are omitted from the training datasets and test datasets.

8.2.2 Layered Knowledge Graph

A knowledge graph is a graph with entities of different types as nodes and several relations between them as edges. Knowledge graph aims to represent entities and the relations between entities. We consider that a Deep Neural Network, which can produce suitable outputs with fewer weak connections between the neurons, would be nearer to a representation of a knowledge graph in a given field. We have selected a fully connected DNN with 2 hidden layers and 12 neurons in each hidden layer for our case study. In this selected neural network, the input layer contains 15 neurons, representing the predefined input features for house cost assessments, and a single output neuron representing the assessed house

value or predicted selling price. We derived a layered knowledge graph for the real estate domain. We can see that the first layer contains 15 input nodes indicating that all selected features have important influences on house price assessments. The total number of hidden neurons in the second and third layer has been reduced from 12 to 10 and 7, respectively. This is due to the weak links and weak neurons being removed from the network. Finally, a single neuron in the last layer represents the assessed house value or predicted selling price

8.2.3 Structured DNN

A deep neural network is a type of neural network in which there exist a certain level of complexity. It is a neural network with more than two layers. The structured DNN is considered to match with the knowledge graph. We made experiments on fully-connected DNNs with different numbers of hidden layers and different numbers of neurons in each hidden layer. The structured Deep Neural Network has four layers such as an input layer, two hidden layers, and an output layer. We set up suitable hyper-parameters for the structured DNN, and trained it using algorithm such as standard feed forward and back propagation algorithm with problem-specific real-time training and fitting techniques. Note that the first layer of the network contains 15 input neurons, which always produce outputs, as there are no biases are connected to the input layer neurons. Even though minor initialized weights create a neural network learn slowly, with sufficient offered data points, adjust a deep neural network with lesser weights which will help to get improved generalization, so to get good performance.

8.3 Component Design

8.3.1 Class Diagram

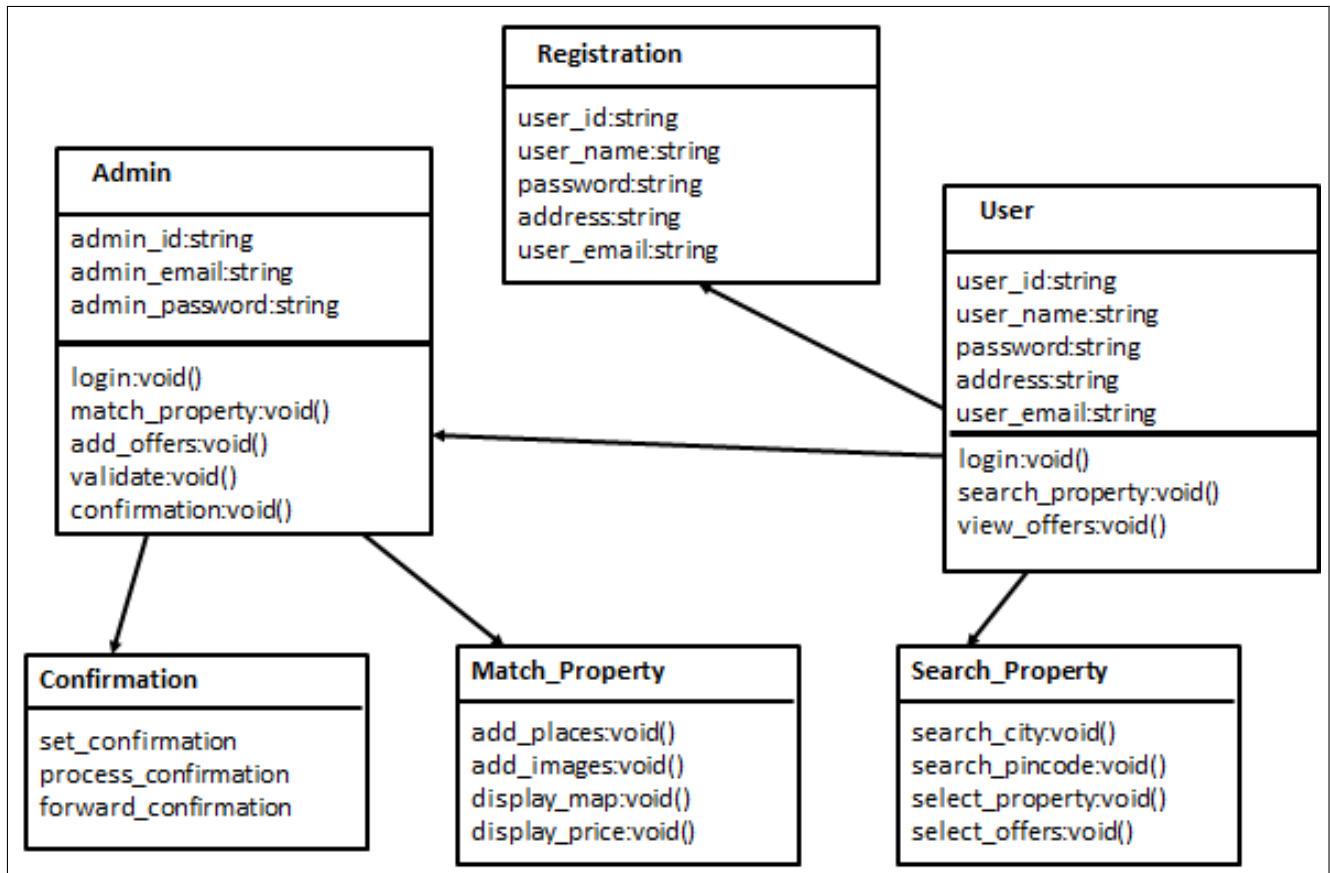


Figure 8.2: Class Diagram

8.3.2 Sequence Diagram

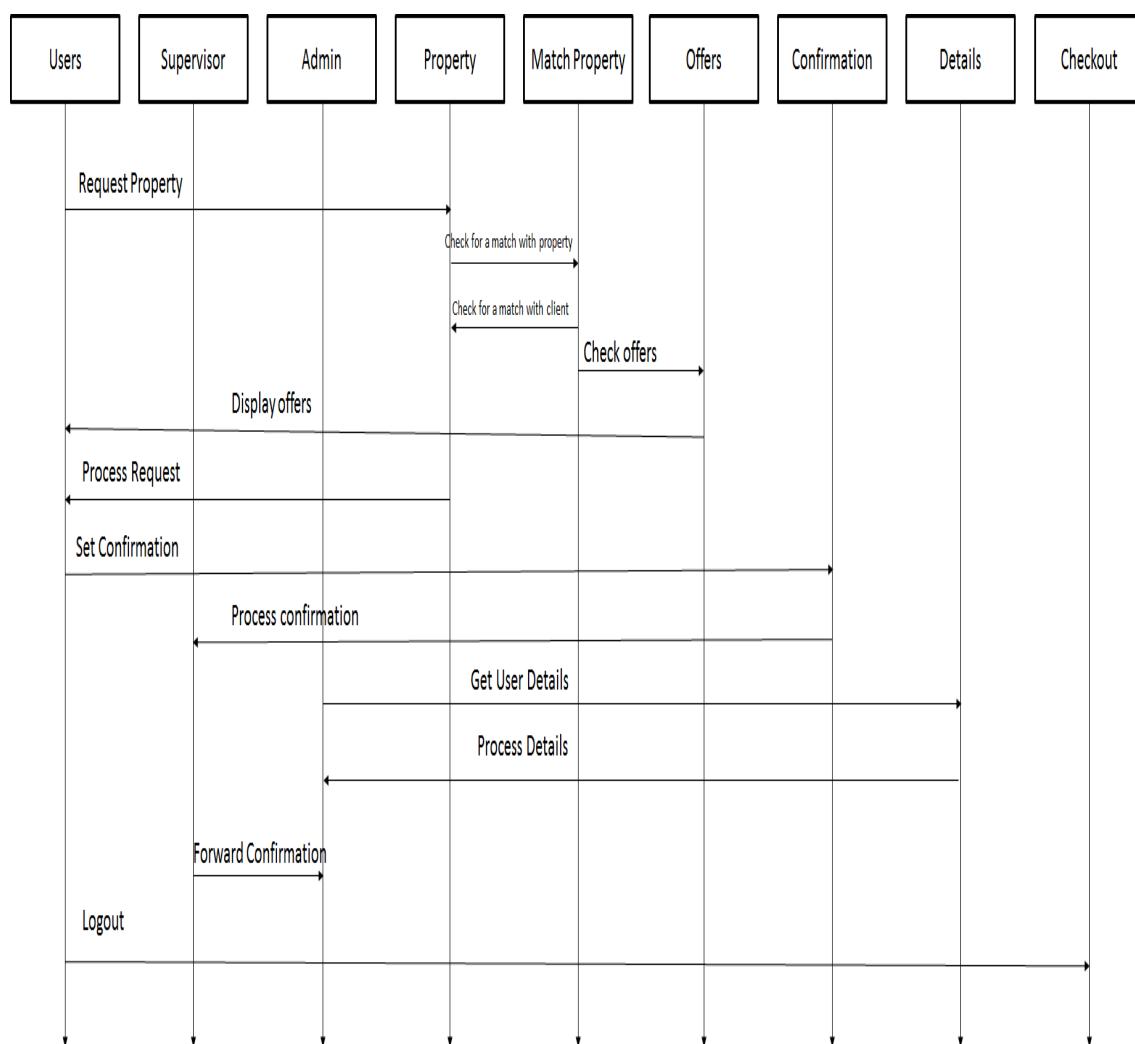


Figure 8.3: Sequence Diagram

8.3.3 Deployment Diagram

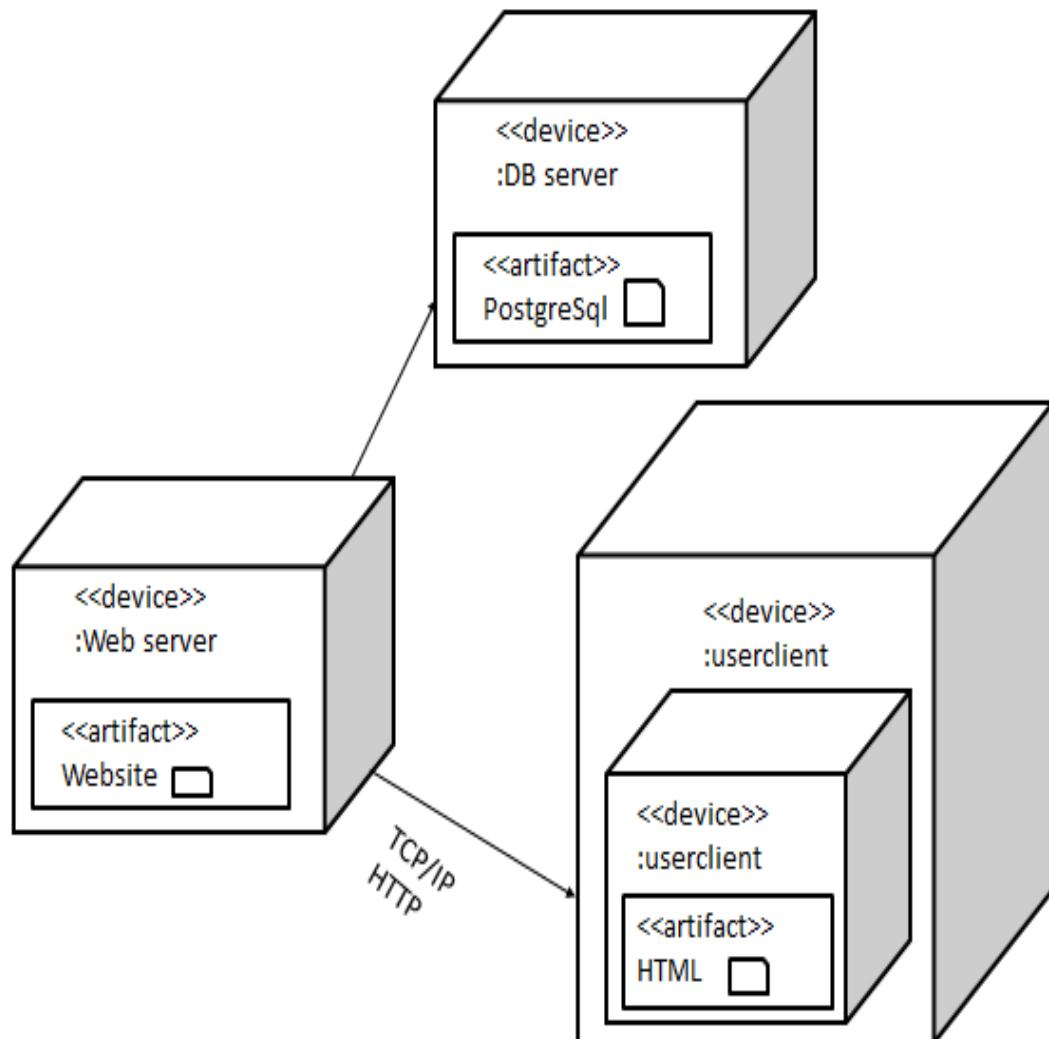
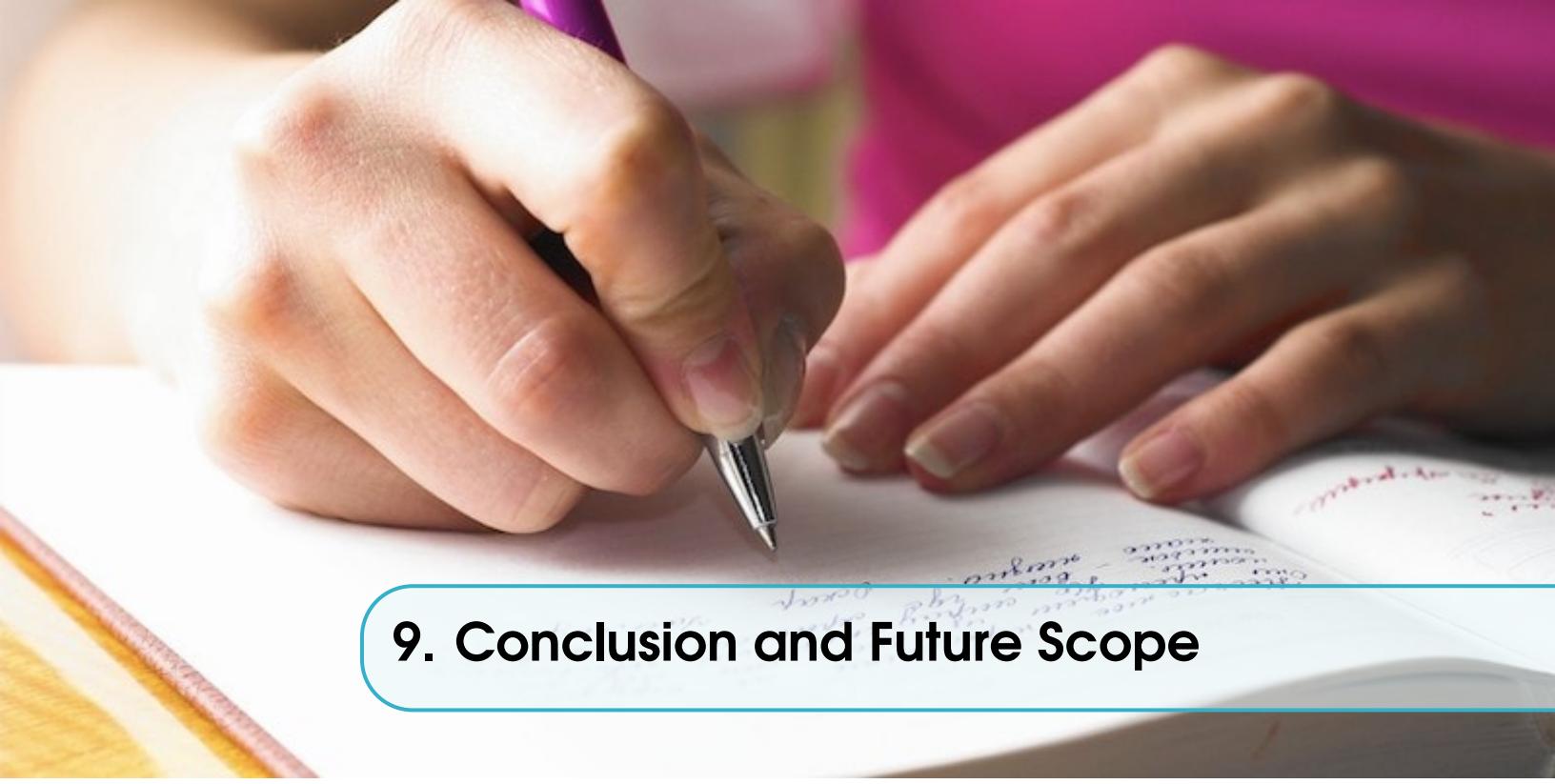


Figure 8.4: Deployment Diagram



9. Conclusion and Future Scope

PREDICTIVE analytic has become more and more important in the world of big data and machine learning. Here for prediction purpose we are going to use Deep Neural network instead of Regression analysis. Due to the large size of a deep learning architecture, deep learning typically requires a large amount of data to train the model, which makes the training process very time and space inefficient. In addition, a large deep architecture may also prone to have the over-fitting issue. Our experimental results given in research paper show that the proposed approach outperforms other conventional methods and leading real estate companies such as *Zillow* and *Redfin*, with significantly improved accuracy for house-price prediction.

For future research, we will study how to automate the process of extracting layered knowledge graphs from the real estate domain based on historical data, and design structured DNNs using the graphs. We will allow a DNN model to automatically change its network structure along the time, so it can be more scalable and better adapt to new market changes. Furthermore, we plan to implement our approach using mobile cloud computing. that supports assessments of real estate via mobile devices with computation functions deployed in the clouds. Finally, we

will try to apply our approach to predictive analytic problems from other domains. such as stock market, health-care, transportation, marketing, e-commerce, security, business, and many more.

References

- (R) Haiping Xu and Amol Gade, “*Smart Real Estate Assessments using Structured Deep Neural Networks*” IEEE 2017, pp.255-266.
- (R) Quanzeng You, Ran Pang, Liangliang Cao, and Jiebo Luo, Fellow, IEEE “*Image Based Appraisal of Real Estate Propertie*” in CIKM, 2017, pp. 481–490.
- (R) N. Nguyen and A. Cripps Predicting Housing Value: “*A Comparison of Multiple Regression Analysis and Artificial Neural Networks*,” Journal of Real Estate Research, vol. 22, no. 3, pp. 313-336, 2016
- (R) Y. E. Hamzaoui and J. A. H. Perez, “*Application of Artificial Neural Networks to Predict the Selling Price in the Real Estate Valuation Process*” in MDM, vol. 2, 2015, pp. 31–36.



Information of Project Group Members

1. Name : Patne Nikhil Sunil
2. Date of Birth : 29/10/1997
3. Gender : Male
4. Permanent Address : A/p Dhom, Tal.-Wai,
Dist.-Satara
5. E-Mail : nikhilpatne94@gmail.com
6. Mobile : 8605358076

1. Name : Shinde Abhishek Sunil
2. Date of Birth : 14/03/1997
3. Gender : Male
4. Permanent Address : A/p Phaltan, Tal.-Phaltan,
Dist.-Satara
5. E-Mail : shindea890@gmail.com
6. Mobile : 7276643616

1. Name : Gholap Satyawan Ashok
2. Date of Birth : 02/10/1996
3. Gender : Male
4. Permanent Address : A/p Varand, Tal.-Daund,
Dist.-Pune
5. E-Mail : satyawanraje7@gmail.com
6. Mobile : 9561522780

1. Name : Dange Namrata Milind
2. Date of Birth : 08/05/1998
3. Gender : Female
4. Permanent Address : A/p Junction, Tal.-Indapur,
Dist.-Pune
5. E-Mail : dangenm@gmail.com
6. Mobile : 8600551630

1. Name : Beera Vivechana Ramesh
2. Date of Birth : 26/01/1995
3. Gender : Female
4. Permanent Address : A/p Baramati, Dist.-
Pune
5. E-Mail : vivechanab5@gmail.com
6. Mobile : 9823838097

©<https://github.com/AbhieShinde/Smart-Real-Estate-Assessment>