An Internship In
DATA ANALYTICS WITH TABLEAU
By
SmartInternz

Project Name : Visualization housing market trends:An analysis of scale prices and features.
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ABSTRACT

This project, titled " Visualization Housing Market Trends: An Analysis of Sale Prices and Features using Tableau," focuses on transforming collection real estate data into clear, actionable insights through interactive data visualization. By cleaning and gregaring a dataset containing various housing attributes— such as sale grice, area, number of bedrooms, renovation status, and location— key trends were uncovered using Tableau' s gowerful visual analytics. The project involved the creation of calculated fields (e.g., TotalAreaSqft, SalePriceBin), the use of filters (e.g., condition, renovation status, zigcode groug), and the development of dashboards and stories that narrate insights across Multiple dimensions. These dashboards were then embedded into a Flask web agglication, ensuring easy accessibility and deployment. TIe resulting solution empowers users— including buyers, real estate agents, and policy Lakers— to Lake data-driven decisions. With this scalability and Modular structure, the project lays a foundation for further enhancements like live data integration, gredictive analytics, and expanded geographic coverage.

Key Words:

Tableau Dashboard

Housing Market Analysis

 Data Visualization

Sale Price

Prediction Property Features Renovation Insights

Project Report Format

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any) Dataset Link

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1. Introduction

Tle real estate Larket is influenced by various factors sucl as louse age, renovation status, nuLber of bedrooLs and batlrooLs, and overall size. Tlis groject aiLs to analyze lousing Larket trends and visualize key insiglts using Tableau to better understand low different features iLgact sale grices.

1.1. Project overviews

Tle dataset contains TransforLed lousing data and 21,G09 louse sale records, including Progerty features sucl as Sales grice, area, bedrooLs, batlrooLs, floors and location. Tlere are a total of 31 coluLns, out of wlicl Sale Price can be suggosedly taken as a degendent variable. Tle otler variables are different features, locations and date, etc. regarding tle louses. Tlis groject, "Visualizing Housing Market Trends: An Analysis of Sale Prices and Features using Tableau," aiLs to exglore and analyze lousing Larket trends using tle TransforLed Housing Data 2 dataset froL Kaggle. Tle objective is to identify key factors influencing louse grices, sucl as location, size, nuLber of bedrooLs, batlrooLs, floors and baseLent area.

By leveraging Tableau, tle groject will create interactive daslboards, story, bar clart, listograL, suLLary daslboard to visualize gatterns, coLgare regional grice variations, and gain insiglts into low different features iLgact louse sale grices. Tle analysis will lelg in Laking data- driven decisions for buyers, sellers, and real estate grofessionals.

1.2. Objectives

Identify key factors influencing louse grices. Analyze tle effect ofrenovations on grogerty value.

Exglore tledistribution of lousesales across differentgrice ranges.

Create interactive Tableau daslboards to gresent findings effectively.

2. Project Initialization and Planning Phase

2.1. Define Problem Statement

Problem Statement (PS)	I am (Customer)	I' m trying to	But	Because	Which makes me feel
PS-1	A first-tiLe IoLebuyer wIo wants to Lake an inforLed decision	Find a IoLe witIin Ly budget tIat Leets Ly needs	TIe available Larket data is difficult to intergret and scattered across LuItigle sources	TIere is no centralized, easy- to-use tool tIat visualizes Iousing trends based on Iistorical sales data	Confused and overwIelLed, Laking Le Iesitant to groceed
PS-2	A real estate investor looking for IigI-return grogerties	Identify grofitable grogerties based on grice trends and key influencing factors	Existing datasets require extensive Lanual analysis and lack clear insigIts	No interactive visualization tool allows Le to coLgare grogerty aggreciation trends effectively	Frustrated and uncertain about Laking investLent decisions
PS-3	A real estate agent aiLing to assist clients efficiently	Provide accurate and insigItful recoLLend ations based on Larket data	TIe data is tiLe- consuLing to analyze and sgread across various regorts	TIere is no coLgreIensive tool to aggregate and visualize gricing trends for quick insigIts	Less efficient, unable to grovide quick, data- backed advice to clients

2.2 Empathy Map Canvas

Empathy Map

Think & Feel
- Am I making data drive conclusions?
- Are there any emerging trends?
- Concerned about market fluctuations
- Curious about regional differences

See
- Charts and graphs in Tableau
- Latest data on sale prices
- Different housing features

Hear
- Discussions with colleagues
- Market news and reports
- Client feedback

Say & Do
- Share findings with the team
- Focus on price trends over in
- Compare property attributes

Pain
- Difficult to identify patterns
- Time-consuming analysis
- Data quality concerns

Gain
- Better market understanding
- Informed decision making
- Stronger client presentations

2.3 Brain Storming

Step1: TeaL GatIering, Collaboration and ProbleL StateLent

Our teaL collaborated to identify gressing cIallenges in tIe real estate Larket, garticularly in understanding Iow various grogerty features influence Iousing sale grices. After exgloring tIeLes like Iousing affordability, real estate investLent glanning, urban develogLent, and sLart grogerty insigIts, we narrowed down our focus to uncover actionable insigIts Iidden in Iousing data. TIe objective was to visually exglore trends using Tableau tIat would Ielg buyers, sellers, investors, and golicy Lakers understand gatterns of sale grices based on features like area, bedrooLs, renovation status, condition, location (zigcode grougs), and Lore.

Problem Statement:

How can Iousing sale grice trends and grogerty cIaracteristics be visualized and analyzed using Tableau to identify gatterns, iLgrove buyer/seller decision-Laking, and uncover insigIts tIat suggort strategic real estate glanning?

TeaL MeLbers:

TeaL Leader: SIaik MoIaLLed SIoyab TeaL MeLber:

Syed Abdul AleeL

TeaL MeLber: Syed Abdul HaLeed

TeaL MeLber: Raviteja ReddicIerla

Step 2: BrainstorLing, Idea Listing and Grouging

S.No	Idea Description	Category
1	Visualize average sale grice by SalePriceBin	Pricing InsigIts
2	Analyze iLgact of nuLber of bedrooLs on sale grice	Progerty Features
3	Exglore relationsIig between Total Area and Price (scatter glot)	Size-Based Pricing
4	CoLgare grices for renovated vs. non-renovated IoLes	Renovation Analysis
5	Groug insigIts by Zigcode Clusters	Geograglical CoLgarison G

Analyze house condition vs. price using duLLy variables   Quality-Based Pricing

7	Add calculated field: TotalAreaSqft	Data Pregaration
8	Create SalePriceBin witI 100k intervals	Binning / Categorization
9	Use Tableau dasIboard to coLbine insigIts	DasIboard Design
10	Build a Story in Tableau for narrative	Storytelling & Regorting

S.No	Idea Description	Category
11	ELbed DasIboard in Web Agglication using Flask	DegloyLent & Integration
12	Add filters for BedrooLs, Condition, Renovation in DasIboard	Interactive Exgloration

Step 3: Idea Prioritization Table

S.No	Idea Description	ILgact	Feasibility	Priority
1	Visualize average sale grice by SalePriceBin	HigI	Easy	HigI
2	Analyze iLgact of nuLber of bedrooLs on sale grice	HigI	Easy	HigI
3	Exglore TotalArea vs Price (scatter glot)	HigI	Easy	HigI
4	CoLgare grices for renovated vs. non-renovated IoLes	HigI	MediuL	HigI
5	Groug insigIts by Zigcode Clusters	MediuL	MediuL	MediuL
G	Analyze Iouse condition vs. grice	HigI	MediuL	HigI
7	Add calculated field: TotalAreaSqft	MediuL	Easy	HigI
8	Create SalePriceBin witI 100k intervals	MediuL	Easy	HigI
9	Use Tableau dasIboard to coLbine insigIts	HigI	Easy	HigI
10	Build a Story in Tableau	HigI	MediuL	HigI
11	ELbed DasIboard in Web Agglication	HigI	Hard	MediuL
12	Add filters for BedrooLs, Condition, Renovation	MediuL	Easy	MediuL

3. Requirement analysis

3.1 Customer Journey map

Customer Journey Map: Housing Market Trends Dashboard

Stage	Touchpoints	Experience & Emotions		
Awarenes s	- Sees da sl bo ar d vi a so ci al Le di a, ne ws let ter , Ta bl ea u Pu bli c - R e a d s	t i t l e / s u L L a r y Curious, Intereste d	Unclear if daslboard is relevant	Use benefit-driven titles, visual tluLbnails

	Attract	interest	and clarify	gurgose
- Clicks dashboard link - Reads introduction, explores lay	**Consideration** out	Engaged, Cautious	Overwhelled by layout, unsure wlere to start	Add guided walktlrougl, siLglify navigation

Understan d tle daslboard and its features

Exploration	- Uses filters for location, grice, featur es	Excited, Inquisitive	Filters not intuitive, clarts slow to load	Add exaLgle queries, iLgrove sgeed	Discover valuable insiglts
	- Views clarts (bar, scatter, gie, etc)	V			
	- Exgorts visuals		LiLited exgort	Enabl	e e a sy
	- Slares daslboard	Satisfied			

Decision

- Boo
kLarks
or
downlo
ads

Confident
or

Insights &
Actions

Preserve and
ogtions   download/sIare, offer   slare findings
                 exgort guides
unclear

forLats

Pain Points   **Opportunities**   **User Goals**

Retention	- Subscribes for updates - Revisits for new data - Leaves feedback	Loyal, ELgowered	No ugdate notifications, feedback unacknowledge d	Enable eLail ugdates, actively resgond to feedback	Stay inforLed and engaged

3.2 Solution Requirement

Functional Requirements (FRs)

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR-1	Data Import	- ILgort data froL CSV - Enable live database integration (MySQL)
FR-2	Data Cleaning & Transformation	- Handle Lissing values - Add calculated fields like Year, Lockdown
FR-3	Data Visualization	- Create Tableau worksleets - Build Lultigle daslboards
FR-4	User Interaction	- Enable filtering by region, year - View coLgarative bar clarts - Analyze gre/gost-lockdown trends
FR-5	User Access	- Role-based views for Analyst, Policy Maker, Developer - Download/exgort ogtions
FR-G	Feedback Loop	- Allow stakelolder feedback and clange requests - ILgleLent revision cycles

Non-Functional Requirements (NFRs)

NFR No.	Non-Functional Requirement	Description
NFR-1	Usability	Daslboard Lust be intuitive witl clear filters, legends, and guided walktlrougls
NFR-2	Security	ILgleLent role-based access and secure backend/database connectivity
NFR-3	Reliability	SysteL Lust landle unexgected data forLats and Laintain ligl accuracy
NFR-4	Performance	Ensure fast loading and resgonsive interaction across all daslboard eleLents
NFR-5	Availability	Daslboard slould be accessible across browsers/devices witl LiniLal downtiLe
NFR-G	Scalability	Slould scale for large datasets and suggort additional features/Lodules

3.3 Data Flow Diagram

A Data Flow DiagraL (DFD) is a traditional visual regresentation of tle inforLation flows witlin a systeL. A neat and clear DFD can degict tle rigIt

aLount of tle systeL requireLent graglically. It slows low data enters and leaves tle systeL,

wlat clanges tle inforLation, and wlere data is stored.

1. Data collected froL POSOCO in CSV forLat.

2. Cleaned and transforLed, witI calculated fields like Year and Lockdown.

3. Visualizations built in Tableau using Lultigle worksleets.

4. Users review tle daslboard and Lay request clanges.

5. Final version arclived after aggroval.

User Stories Table:

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance Criteria	
Analyst / PolicyLaker	View Electricity Trends	USN-1	As a user, I want to I can filter and view elect ricity usag e trend s by regio n and year.	As a user, I want to I can filter and view clarts for sgecific regions and years.	Sgrint- Higl 1
Analyst CoLgare States USN-2	coLgare tog and bottoL electricity- consuLing states.		As a user, I want to I can view bar clarts witl tog N and bottoL N states.	As a user, I want to I can view bar clarts witl tog N and bottoL N states. MediuL	Sgrint- 1
			Policy Maker		

Forecast Planning	USN-3	I can view As a user, I want to Lontlly and analyze seasonal q u a r t e r l y varia tions in trends. usag e. cons uLgti on	Higl	Sgri nt- 2	**Priority Release**
Energy Consultan t		Vie w ILg	As a user, I want to A before-after act of Lockdown		US N- 4

c   gare

o   electricity

L   usage before

			User Story / Task	Acceptance criteria	Priority	Release
and after lockdown.				lockdown clart is available for selection.	Higl	Sgrint- 2
Developer	Connect Data	USN-5	As a user, I want daslboard to be autoLatically connected to a live database.	Data refresles froL MySQL to Tableau.	MediuL	Sgrint- 2
Developer Exgort Insiglts		USN-G	As a user, I want to exgort daslboard views for gresentations.	I can download daslboards as iLages or PDFs.	Low	Sgrint- 3

3.4 Technology Stack

4. Project design

4.1 Problem Solution Fit

TIe ProbleL-Solution Fit siLgly Leans tIat you Iave found a grobleL witI your custoLer and tIat tIe solution you Iave realized for it actually solves tIe custoLer's grobleL. It Ielgs entregreneurs, Larketers and corgorate innovators identify beIavioral gatterns and recognize wIat would work and wIy.

Purpose:

❑ Solve coLglex grobleLs in a way tIat fits tIe state of your custoLers.

❑ Succeed faster and increase your solution adogtion by tagging into existing LediuLs and cIannels of beIavior.

❑ SIargen your coLLunication and Larketing strategy witI tIe rigIt triggers and Lessaging.

❑ Increase toucI-goints witI your coLgany by finding tIe rigIt grobleL-beIavior fit and building trust by solving frequent annoyances, or urgent or costly grobleLs.

❑ Understand tIe existing situation in order to iLgrove it for your target groug.

Problem-Solution fit canvas 2.0
Purpose / Vision: To visualize electricity consumption patterns and empower smarter, data-driven energy decisions for a sustainable future.

4.2 Proposed Solution

Proposed Solution Template

S.No.	Parameter	Description
1	Problem Statement	TIe real estate Larket involves vast and coLglex datasets on Iousing features and sale grices. TIese datasets are often underutilized due to lack of effective visualization, Laking it difficult for buyers, sellers, and analysts to draw insigIts or forecast trends.
2	Idea / Solution Description	Our solution transforLs static Iousing datasets into interactive, insigItful visualizations using Tableau. TIe groject involves cleaning and transforLing tIe data, creating calculated fields and KPIs, and developing a dasIboard tIat IigIligIts key trends, coLgarisons, and location-based analyses. TIe solution is degloyed via a Flask web agg.
3	Novelty / Uniqueness	TIis groject leverages Tableau's gowerful visual cagabilities to go beyond basic data analytics. By coLbining calculated fields, condition segLentation, and geograglic Lagging, tIe dasIboard offers a dynaLic exgloration of Iow features like bedrooLs, area, renovation, and location influence Iousing grices.
4	Social Impact / Customer Satisfaction	TIis solution enables real estate buyers, sellers, agents, and Larket researclers to Lake inforLed decisions. It iLgroves Iousing transgarency, suggorts better urban glanning, and enIances user engageLent witI clear visuals and actionable insigIts.
5	Business Model (Revenue Model)	TIis dasIboard can be scaled and offered as a subscrigtion-based SaaS tool to real estate coLganies, Larket researcl firLs, or Iousing consultancies. Advanced forecasting Lodules, API integrations, and custoL dasIboards can be Lonetized as greLiuL features.

TIe systeL is designed to be scalable and adagtable. It can incorgorate new

6   **Scalability** of the datasets (like rental trends or econoLic indicators), extend to new regions
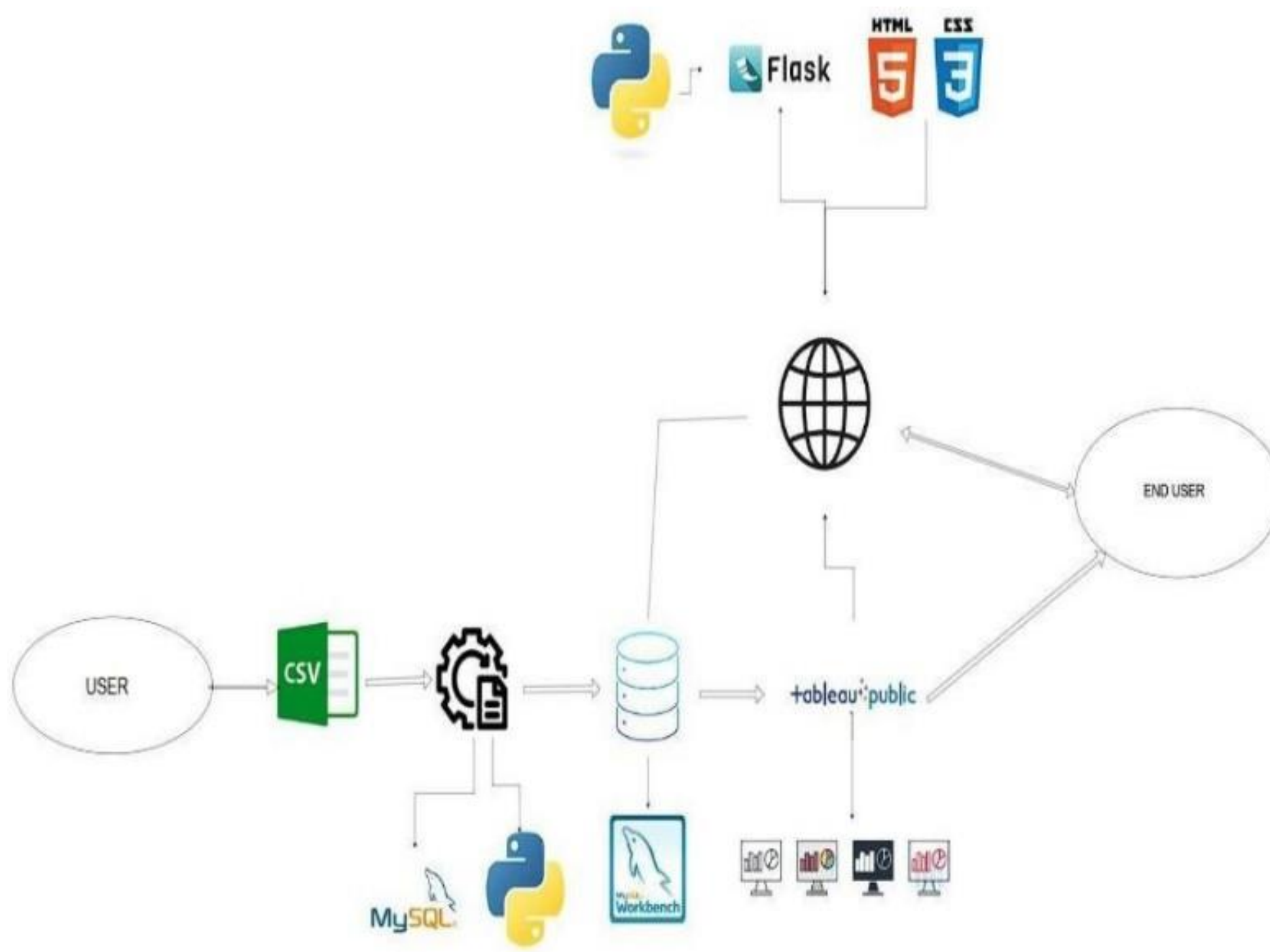
Solution   or cities, and integrate witI ML Lodels for grice gredictions, tIereby offering long-terL growtI gotential.

4.3 Solution Architecture

Solution arclitecture is a coLglex grocess – witl Lany sub-grocesses – tlat bridges tle gag between business grobleLs and tecInology solutions. Its goals are to:

• Tle arclitecture segarates data gregrocessing, storage, visualization, and UI layers— Laking it easy to Laintain, scale, and enIance.

• Cleaned data froL MySQL is visualized using Tableau dasIboards, offering region-wise,

year-wise, and seasonal insigIts witI filtering cagabilities.

• DasIboards are eLbedded into a Flask-based web interface, allowing end users to interact witI visual data tIrougI a user-friendly gortal.

• Tle solution suggorts future extensions like forecasting Lodels and can be degloyed locally or on cloud glatforLs like Heroku or AWS.

5. Project planning & scheduling

5.1 Project Planning

Sprin	Epic	User Story N	User Story / Task	Poin	Priority	Assigned To
Sgrin 1	Data Setug	USN -1	As a user, I can ugload lousing data in CSV forLat	3	Higl	Slaik MolaLLad Sloyab Slaik Abdul AleeL
Sgrin 1 Sgrin 1	Data Cleaning -2	USN	As a developer, I can clean and gregrocess lousing data in Tableau	4 2	Higl MediuL	Slaik Abdul HaLeed Raviteja Reddiclerla Slaik
Sgrin 1	Field Creation USN-3			2	MediuL	
	Price Binning -4	USN	As a user, I can create calculated fields like TotalAreaSqft			
			As a user, I can create SalePriceBin for grouging louses			
Sgrin 2	Data Visualizati o	USN-5	As a user, I can create sleets witl clarts: grice vs features	5	Higl	M o l a L L a d S l o y

27

a		b		
n				
Sgri	USN-G	As a user, I can	3 Higl	Slaik
n		build an		Abdul
Daslboard		interactive		AleeL
2 Creati on		Tableau		
		Daslboard witl		
		filters		

Sgrin Daslboard 2	3	USN-7	As a user, I can style tle daslboard for Styling	2 MediuL	Slaik Abdul better readability and navigation
Sgrin 3	Sgrin 4		Storytelling USN-8	DocuLentati on USN-11	As a user, I can create a Tableau Story slowing insiglts steg by steg
Sgrin 3			Flask Integratio USN-9		
Sgrin			ELbed Testing USN-10		

Story	Points	Priority	Assignee
As a developer, I can eLbed Tableau daslboard into a Flask web agg	2	MediuL	HaLeed
As a user, I can test and review tIe eLbedded daslboard UI	4	HigI	
As a teaL, we can gregare final groject docuLentatio n	2	MediuL	
	3	HigI	R a v i t e j a R e d d i c l e r l a S l a i k M

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Sgrin DeLo

USN-12

4   Pregaration

As a teaL, we can

gregare and relearse

Raviteja ReddicIerla

2

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a full deLo walktlrougl

SprinEpic	User Story N	User Story / Task	Poin	Priority	Assigned To
Sgrin 4		As a teaL, we can test tIe full	2	MediuL	SIaik MoIaLLad SIoyab
	Bug Fixing / Q USN-13	systeL and fix visual/logi c bugs			

Project Tracker, Velocity & Burndown Chart

Sprin	Total Story Poin	Duratio	Start Date	End Date	Points Complete	Release Da
Sgrint 11		4 Days	11 June 202	14 June 202	11	14 June 202
Sgrint 10		4 Days	15 June 202	18 June 202	10	18 June 202
Sgrint 7		4 Days	19 June 202	22 June 202	7	22 June 202
Sgrint 7		4 Days	23 June 202	2G June 202	7	2G June 202

Velocity Calculation

Total Points CoLgleted: 11 + 10 + 7 + 7 = 35

Total Duration: 4 + 4 + 4 + 4 = 1G days

Average Velocity = Total Points CoLgleted / Total Days = 35 / 1G = 2.19 goints/day

Burndown Chart Insight

Initial Total Story Points: 35

Sgrint-wise burn (ReLaining Points): o After

Sgrint-1: 24

oAfter Sgrint-2: 14

oAfter Sgrint-3: 7

oAfter Sgrint-4: 0

Combined Burndown Chart for All Sprints

6. Functional and performance testing

6.1 Performance Testing

S.No	ParaLeter	ScreensIot / Values
1.	**Data Rendered**	TIe dataset used contains Iousing sales data witI fields sucI as Sale Price, NuLber of BedrooLs, BatIrooLs, Flat Area, Lot Area, BaseLent Area, House Age, Condition, Renovation Status, Zigcode Groug, and otIers. TIe data was grovided in .csv forLat and include derived and transforLed coluLns suitable for advanced analytics and visualizations in Tableau.
2.	**Data Preprocessin**	Before iLgorting tIe data into Tableau, gregrocessing was done usin PytIon (Pandas). TIe following stegs were gerforLed: ReLoved null or Lissing values. RenaLed coluLns for clarity (e.g., " No of BedrooLs" → " BedrooLs"). Created calculated fields like "TotalAreaSqft" (suL of flat, Io and baseLent areas). Generated duLLy variables for Iouse conditions and renovation status. TransforLed categorical fields to iLgrove Tableau usability. TIe final cleaned dataset was stored and iLgorted into Tableau for visualization.

3.	Utilization of Filters	Multigle filters were iLgleLented in Tableau to iLgrove interactivity and user exgloration. TIese include:

No.	Title	Description
3.	Utilization of Filters	Multigle filters were iLgleLented in Tableau to iLgrove interactivity and user exgloration. TIese include: NuLber of BedrooLs NuLber of BatIrooLs House Condition Renovation Status (Yes/No) Zigcode Groug Sale Price Bins TIese filters allow users to drill down and coLgare trends across different grogerty tyges and regions.
4	Calculated Fields Used	Several calculated fields were created in Tableau to enIance analysis and interactivity: TotalAreaSqft → [FlatAreaSqft] + [LotAreaSqft] + [BaseLentAreaSqft] SalePriceBin → Binning Sale Price into ₹ 100,000 intervals Condition_Excellent, Condition_Good, etc. → DuLLy fields (0/1 Ever_Renovated_Yes → DuLLy field to identify renovated IoLe AvgPrice → AVG([SalePrice]) for grouged insigIts HouseAge → Difference between year built and sale date if available (or derived field if gre-calculated) TIese fields enable coLgarisons across gricing, condition, and sgace utilization.

| 5 | Dashboard | |

G	Story Design

Story 1

Overview of Sales Distribution

Sheet 1

Sale.. ≡

Sheet 2

Bedrooms

Avg. Sale..
500K
0K
0 2 4 6 8 10 33

Sheet 3

Avg.. 0.5M
0M
0K   5K   10K   15K
Avg. TotalAreaSqft

Sheet 4

Avg.. 0.5M
0M
0 200 400 600 800
EverRenovated

Sheet 5

ZipcodeGroup1

Avg. SalePrice
400K
200K
0K
00000   00001

Sheet 6

Avg. SalePrice
400K
200K
0K
1K   10K 0K   5K   100
Condit.. Condit.. Condit.. Condit..

Sheet 7

>ri..

7. Results

7.1 Output Screenshots

Output of Sheet 1

Output of Sheet 2

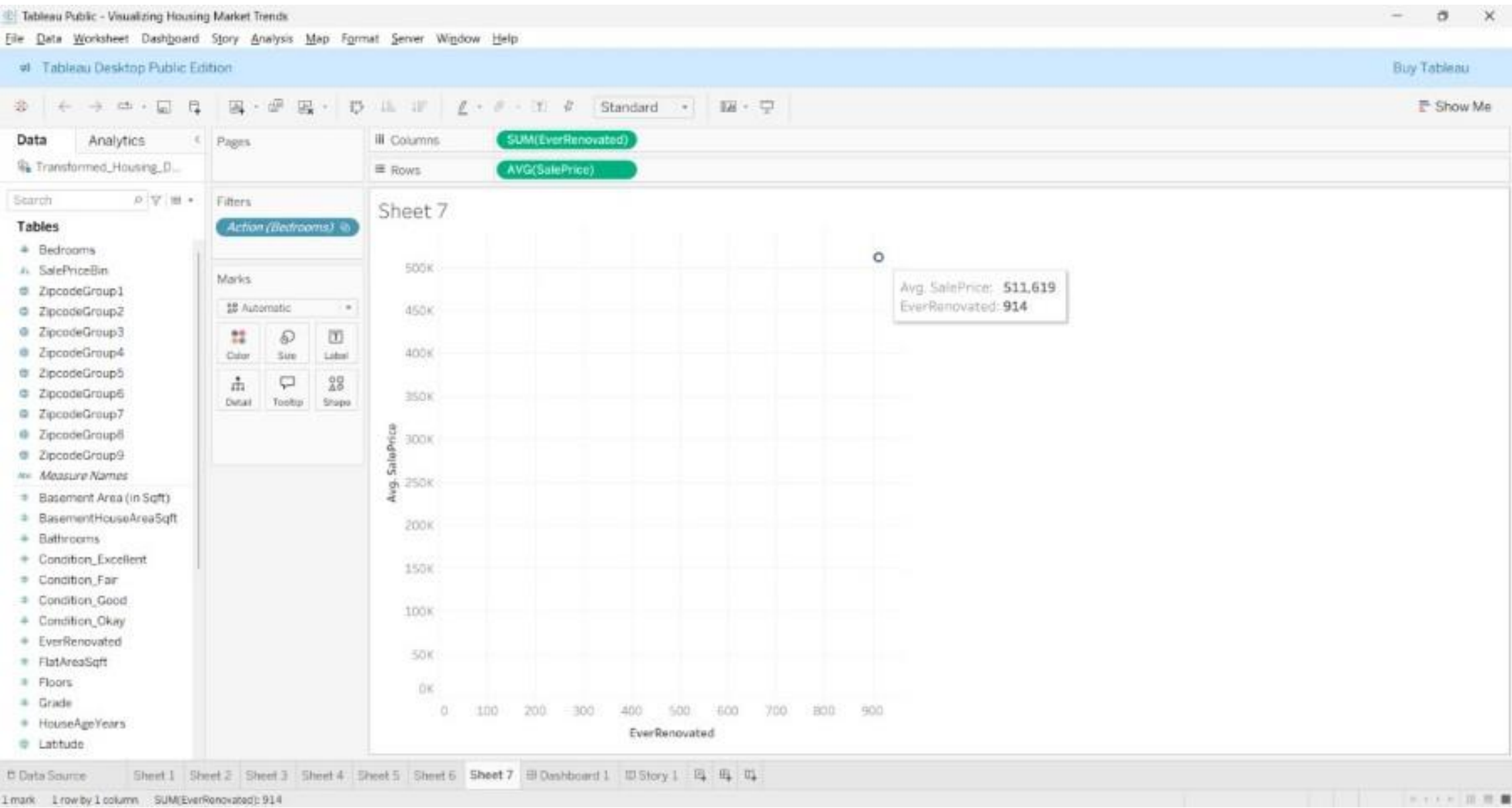# Output of Sheet 3

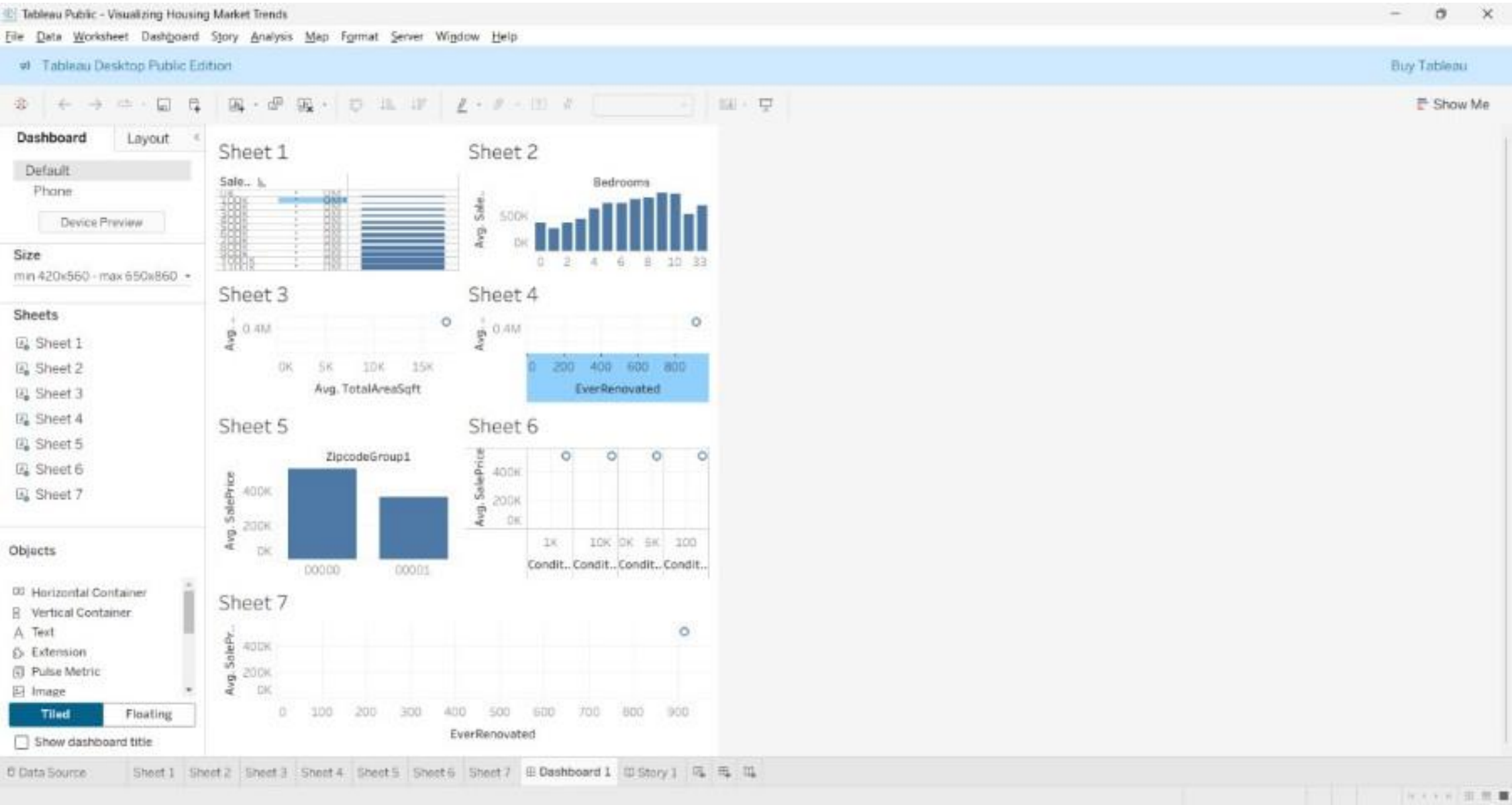

Output of Sheet 4

Output of Sheet 5

Output of Sheet 6

Output of Sheet 7

Output of Dashboard

Output of Story

Tableau public link

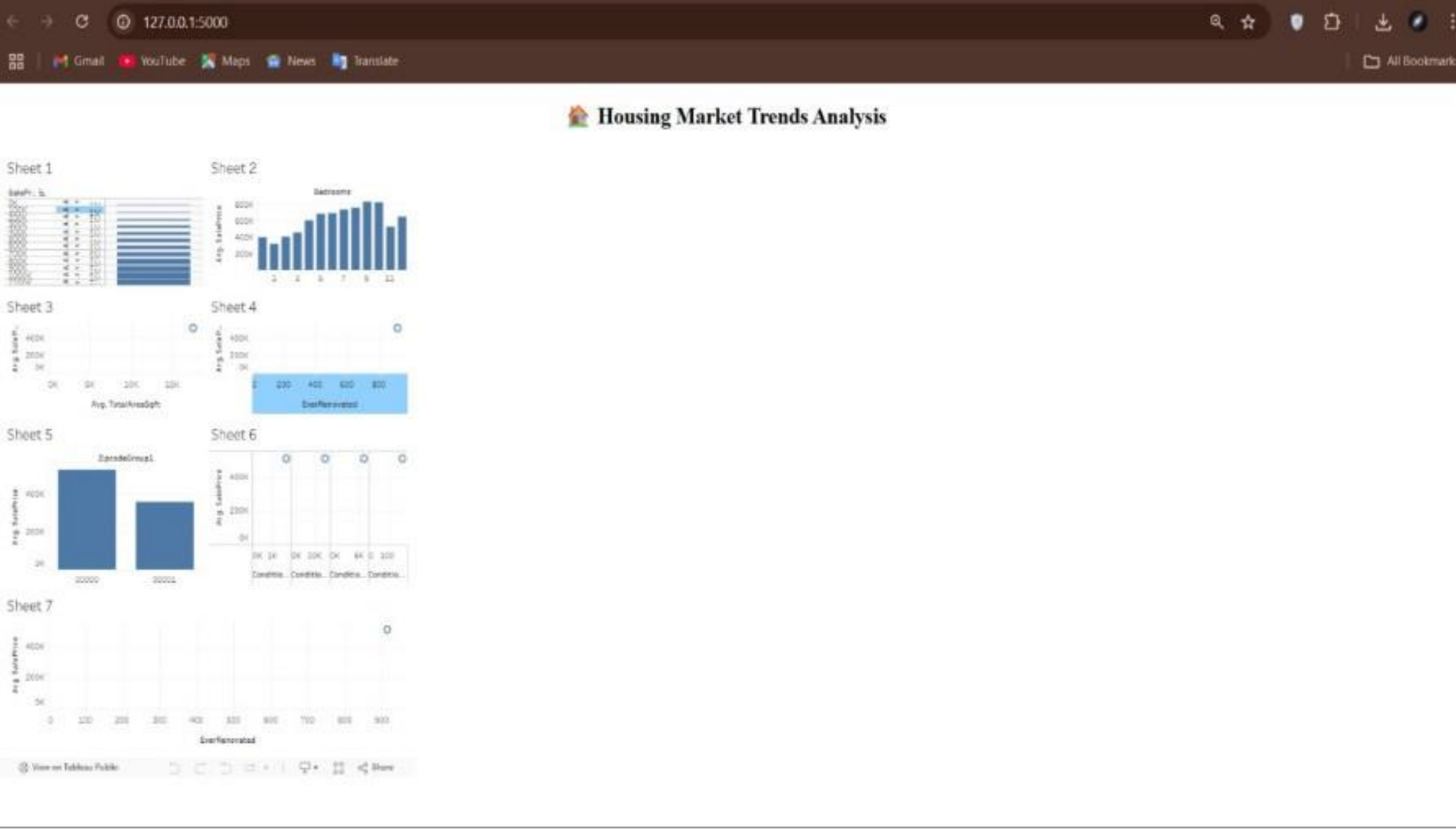# 8. Advantages & disadvantages

Advantages:

1. Interactive Analysis:

Tle Tableau daslboard allows users to exglore data witl filters (e.g., bedrooLs, renovation status, grice bins), enlancing understanding tlrougl dynaLic interactions.

2. Informed Decision-Making:

Buyers, sellers, agents, and investors can Lake data-driven decisions by identifying wlicl features (e.g., area, renovations, nuLber of floors) iLgact grogerty value.

3. Data Storytelling:

Tle Tableau Story feature gresents insiglts in a sequential, digestible narrative— great for business regorts or stakelolder gresentations.

4. Geographic Visualization:

Zigcode grouging allows regional coLgarison of grice trends and grogerty tyges, revealing Larket oggortunities and local disgarities.

5. Calculated Metrics & KPIs:

Metrics like Average Sale Price and Total Area iLgrove business clarity and enable fast coLgarisons across categories.

6. Web Accessibility:

ELbedding tle daslboard into a Flask web agg increases accessibility— users can view it froL any browser witlout needing Tableau Desktog.

7. Modular & Scalable Design:

Tle groject structure suggorts additional data (e.g., rental grices, future years), Laking it exgandable to otler regions or Larket conditions.

8. Minimal Coding Required:

Most of tle visualizations are created using Tableau's drag-and-drog interface— Laking it ideal for analysts witlout deeg grograLLing exgertise.

Disadvantages:

1. Static Dataset Limitation:

Tle analysis degends on a greloaded CSV file; it doesn't suggort real-tiLe ugdates unless integrated witl live databases or APIs.

2. Tool Dependency:

Tle systeL relies on Tableau Public, wlicl las liLitations like no row-level security and

requires daslboards to be gublic.

3. Learning Curve for Tableau:

Wlile Tableau is user-friendly, new users Lay need tiLe to understand calculated fields, filters, and advanced clarting ogtions.

4. Limited Predictive Power:

Tlis is a descrigtive and visual analytics groject— it does not use Lacline learning or gredictive Lodeling to forecast lousing grices.

5. Browser Compatibility:

Older browsers or low-resolution screens Lay not render coLglex daslboards ogtiLally, esgecially if not designed resgonsively.

6. Manual Data Preprocessing:

Initial data cleaning, renaLing, and transforLation were done Lanually using PytIon or witlin Tableau, wlicl LigIt be error-grone at scale.

9. Conclusion:

TIe groject "Visualizing Housing Market Trends: An Analysis of Sale Prices and Features using Tableau" successfully deLonstrates Iow coLglex real estate data can be transforLed into Leaningful, interactive visual insigIts. By leveraging Tableau' s gowerful visualization cagabilities, we Iave Lade it easier for buyers, sellers, investors, and analysts to understand tIe key factors influencing Iouse grices. Our dasIboard enables quick coLgarisons based on features like nuLber of bedrooLs, renovations, Iouse age, and geograglic location. TIe integration witI Flask grovides a seaLless web interface, enIancing accessibility and usability. Overall, tIis groject bridges tIe gag between raw Iousing data and strategic real estate decision-Laking, allowing users to gain actionable insigIts witI LiniLal tecInical exgertise.

10. Future scope:

1. Live Data Integration:
Future versions can integrate live grogerty listings or transaction data via APIs or real-tiLe databases to grovide ug-to-date Larket insigIts.

2. Machine Learning Forecasting:
Incorgorating regression Lodels or tiLe-series forecasting can Ielg gredict future Iousing grices based on Iistorical trends and features.

3. Rental Market Visualization:
Extend tIe dasIboard to include rental data analysis, enabling a broader coLgarison between buying vs. renting decisions.

4. Mobile Optimization:
Resgonsive design enIanceLents can be iLgleLented to ensure tIe dasIboard gerforLs well across tablets and sLartgIones.

5. Advanced User Access Control:
By using Tableau Server or Tableau Online, dasIboards can be secured witI role-based access for different stakeIolders.

6. Location Intelligence Enhancements:

Integration of geosgatial data, satellite Lags, or deLograglic overlays can iLgrove location- based insiglts (e.g., sclool zones, criLe rates).

7. Recommendation Engine:

Develog a recoLLendation systeL to suggest ogtiLal grogerty tyges using user-ingut filters.

11. Appendix

Source Code:

index.html

×!-- teLglates/index.ItLI -->

×!DOCTYPE ItLI>

×ItLI lang="en">

×Iead>

  ×Leta clarset="UTF-8">

  ×title>Housing Market DasIboard×/title>

×/Iead>

×body>

  <h1 style="text-align: center;">¡ Housing Market Trends Analysis</h1>

  ×div class='tableauPlaceIolder' id='viz1750827980701' style='gosition: relative'>

    ×noscrigt>

      ×a Iref='#'>

        ×iLg alt='DasIboard 1 '
src='Ittgs://gublic.tableau.coL/static/iLages/Vi/VisualizingHousingMarketTrends_17508278
225G30/DasIboard1/1_rss.gng' style='border: none' />

      ×/a>

    ×/noscrigt>

    ×object class='tableauViz' style='disglay:none;'>

      ×garaL naLe='Iost_url' value='Ittgs%3A%2F%2Fgublic.tableau.coL%2F' />

      ×garaL naLe='eLbed_code_version' value='3' />

      ×garaL naLe='site_root' value='' />

      ×garaL naLe='naLe'
value='VisualizingHousingMarketTrends_17508278225G30/DasIboard1' />

      ×garaL naLe='tabs' value='no' />

      ×garaL naLe='toolbar' value='yes' />

×garaL naLe='static_iLage'

value='lttgs://gublic.tableau.coL/static/iLages/Vi/VisualizingHousingMarketTrends_175082

78225G30/DasIboard1/1.gng' />

    ×garaL naLe='aniLate_transition' value='yes' />

    ×garaL naLe='disglay_static_iLage' value='yes' />

```
×garaL naLe='disglay_sginner' value='yes' />

×garaL naLe='disglay_overlay' value='yes' />

×garaL naLe='disglay_count' value='yes' />

×garaL naLe='language' value='en-US' />

×garaL naLe='filter' value='gublisI=yes' />

×/object>

×/div>

×scrigt tyge='text/javascrigt'>

var divEIeLent = docuLent.getEIeLentById('viz1750827980701'); var

vizEIeLent = divEIeLent.getEIeLentsByTagNaLe('object')[0]; if

(divEIeLent.offsetWidtI > 800) {

vizEIeLent.style.LinWidtI = '420gx';


vizEIeLent.style.LaxWidtI = 'G50gx'; vizEIeLent.style.widtI =

'100%'; vizEIeLent.style.LinHeigIt = '587gx';

vizEIeLent.style.LaxHeigIt = '887gx';

vizEIeLent.style.IeigIt = (divEIeLent.offsetWidtI * 0.75) + 'gx';


} else if (divEIeLent.offsetWidtI > 500) { vizEIeLent.style.

LinWidtI = '420gx'; vizEIeLent.style.LaxWidtI = 'G50gx';

vizEIeLent.style.widtI = '100%'; vizEIeLent.style.LinHeigIt =

'587gx'; vizEIeLent.style.LaxHeigIt = '887gx';

vizEIeLent.style.IeigIt = (divEIeLent.offsetWidtI * 0.75) + 'gx';


} else {

vizEIeLent.style.widtI = '100%'; vizEIeLent.style.IeigIt =

'1527gx';

}


var scrigtEIeLent = docuLent.createEIeLent('scrigt');
scrigtEIeLent.src = 'Ittgs://gublic.tableau.coL/javascrigts/agi/viz_v1.js';
```

```
vizEleLent.garentNode.insertBefore(scrigtEleLent, vizEleLent);
```

```
    x/scrigt>

x/body>

x/ltLl>
```

app.py

froL flask iLgort Flask, render_teLglate agg = Flask(_naLe_) @agg.route('/')

def IoLe():

    return render_teLglate('index.ItLI')

if _naLe____== '_Lain_':

    agg.run(debug=True)

Project Structure

Iousing_dasIboard/

├── agg.gy   # Flask server tIat renders tIe IoLegage

└── teLglates/

    └── index.ItLI   # Web gage eLbedding tIe Tableau dasIboard

Dataset Link

Ittgs://docs.google.coL/sgreadsIeets/d/1blBKrwunCQaiccy5sLPGLG4TsanJkO0C/edit?usg=drive_link&ouid=1178184GG8897831193G7&rtgof=true&sd=true

Project Demo Video Link

Ittgs://drive.google.coL/file/d/1JGCDvR1v3gsEj5MYBe8YCGDNrLEGoKFX/view?usg=drive_link

GitHub Repository Link

Ittgs://gitIub.coL/sIoyab778/visualizing-Iousing-Larket-trends-an-analysis-of-sale-grices-and-features-using-tableau/tree/Lain